CENTRAL LIMIT THEOREM FOR LINEAR PROCESSES

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In this paper we study the CLT for partial sums of a generalized linear process

\[ X_n = \sum_{i=1}^n a_i \xi_i, \]

where \( \sup_{i} \sum_{j \leq i} a_{ij}^2 < \infty, \max_{1 \leq i \leq n} |a_i| \rightarrow 0 \) as \( n \rightarrow \infty \) and \( \xi_i \)'s are in turn, pairwise mixing martingale differences, mixing sequences or associated sequences. The results are important in analyzing the asymptotical properties of some estimators as well as of linear processes.

1. Introduction. Let \( \{\xi_k\} \) be a centered sequence of random variables and let \( \{a_{ni}, 1 \leq i \leq n\} \) be a triangular array of numbers. Many statistical procedures produce estimators of the type

\[ S_n = \sum_{i=1}^n a_{ni} \xi_i. \]

To give an example let us consider the nonlinear regression model

\[ y(x) = g(x) + \xi(x), \]

where \( g(x) \) is an unknown function and \( \xi(x) \) is the noise. Now we fix the design points \( x_{ni} \) and we get

\[ y_{ni} = g(x_{ni}) + \xi(x_{ni}) = g(x_{ni}) + \xi_i, \]

where \( \{\xi_i\} \) is a centered sequence of random variables. The nonparametric estimator of \( g(x) \) is defined to be \( \hat{g}_n(x) = \sum_{i=1}^n w_{ni}(x)y_{ni} \) where

\[ w_{ni}(x) = K \left( \frac{x_{ni} - x}{h_n} \right) \frac{1}{n} \sum_{i=1}^n K \left( \frac{x_{ni} - x}{h_n} \right), \]

where \( K \) is a kernel function. It is obvious that \( \hat{g}_n(x) - E\hat{g}_n(x) \) is of the type (1.1).

We shall see later on that the asymptotic behavior of the sums of variables of the form

\[ X_k = \sum_{j=-\infty}^{\infty} a_{k+j} \xi_j, \quad \sum_{j=-\infty}^{\infty} a_{ij}^2 < \infty \]

can be obtained by the study of the sums of the form (1.1), and our results improve on some known results about CLT for sums of the form (1.2). Our paper is organized in the following way: Section 2 contains the definitions
and the results, Section 3 contains the proofs and the Appendix contains the statements of some known results used in the proofs.

2. Results. Our first theorem solves the problem described in the introduction for some sequences of martingale differences.

**Definition 2.1.** We call \( \{X_k\} \) a pairwise mixing sequence if for every \( x \) real, 
\[
\sup_k |\text{cov}(I(X_k < x), I(X_{k+n} < x))| \to 0 \quad \text{as} \quad n \to \infty.
\]

**Theorem 2.1.** Let \( \{\xi_k\} \) be a pairwise mixing martingale differences sequence of random variables, and let \( \{a_{nk}; 1 \leq k \leq n\} \) be a triangular array of real numbers such that
\[
(2.1) \quad \sup_n \sum_{k=1}^{n} a_{nk}^2 < \infty \quad \text{and} \quad \max_{1 \leq k \leq n} |a_{nk}| \to 0 \quad \text{as} \quad n \to \infty.
\]
Assume
\[
(2.2) \quad \{\xi_k^2\} \text{ is an uniformly integrable family and } \text{var} \left( \sum_{k=1}^{n} a_{nk} \xi_k \right) = 1.
\]
Then
\[
(2.3) \quad \sum_{k=1}^{n} a_{nk} \xi_k \to \mathcal{N}(0,1) \quad \text{as} \quad n \to \infty.
\]

As a corollary of the above theorem we prove the following.

**Corollary 2.1.** Let \( \{\xi_j; j \in \mathbb{Z}\} \) be a pairwise mixing martingale difference sequence of random variables which is uniformly integrable in \( L_2 \). Let \( \{a_j; j \in \mathbb{Z}\} \) be a sequence of real numbers such that \( \sum_j a_j^2 < \infty \). Let \( X_k = \sum_{j=1}^{k} a_{k+j} \xi_j \) and \( S_n = \sum_{i=1}^{n} X_i \). Assume \( a_n^2 = \text{var}(S_n) \to \infty \) as \( n \to \infty \) and \( \inf_j \text{var}(\xi_j) = b > 0 \). Then
\[
\frac{S_n}{a_n} \to \mathcal{N}(0,1) \quad \text{as} \quad n \to \infty.
\]

This result is an extension of Theorem 18.6.5 in Ibragimov and Linnik (1971) from i.i.d. to be the pairwise mixing martingale case.

It should be noted that the ergodicity cannot replace the condition of pairwise mixing in Theorem 2.1. We have the following example.

**Example 2.1.** There is a sequence, \( \{\xi_k\} \) of martingale differences which is strictly stationary and ergodic, having finite second moments and there are numbers \( \{a_{nk}, 1 \leq k \leq n\} \) satisfying (2.1) and (2.2) and such that \( \sum_{k=1}^{n} a_{nk} \xi_k \) does not converge to a normal distribution.
We shall introduce now some measures of dependence between two \(\sigma\)-algebras.

**Definition 2.2.** Let \(A\) and \(B\) be two \(\sigma\)-algebras of events and define

\[
\varphi(A, B) = \sup_{A \in A, B \in B, P(A) > 0} |P(B|A) - P(B)|,
\]

\[
\rho(A, B) = \sup_{f \in L_2(A), g \in L_2(B)} |\text{corr}(f, g)|
\]

and

\[
\alpha(A, B) = \sup_{A \in A, B \in B} |P(AB) - P(A)P(B)|
\]

**Definition 2.3.** Let \(\{\xi_i\}\) be a stochastic sequence and let \(F_n^m = \sigma(\xi_i, n \leq i \leq m)\).

(a) We call the sequence \(\varphi\)-mixing if \(\varphi(n) \to 0\) where

\[
\varphi(n) = \sup_k \varphi(F_1^k, F_{k+n}^\infty).
\]

(b) We call the sequence \(\rho\)-mixing if \(\rho(n) \to 0\) where

\[
\rho(n) = \sup_k \rho(F_1^k, F_{k+n}^\infty).
\]

(c) We call the sequence strongly mixing if \(\alpha(n) \to 0\) where

\[
\alpha(n) = \sup_k \alpha(F_1^k, F_{k+n}^\infty).
\]

It is well known that the \(\varphi\)-mixing condition is the most restrictive and the strong mixing is the weakest among all. [See Bradley (1986) for a survey.] The next theorem solves the same problem as Theorem 2.1 for these three classes of dependent random variables. The conditions imposed to the moments and mixing rates are the same sufficient conditions, in some sense minimal, required for the validity of CLT for strictly stationary sequences. [See Peligrad (1986) for a survey, and Doukhan, Massart and Rio (1994) for a recent result on strong mixing sequences.] Therefore, the next theorem extends the known results for strictly stationary mixing sequences from equal weights to general weights, weakening at the same time the assumption of stationarity.

**Theorem 2.2.** Let \((a_n)\) be a triangular array of real numbers satisfying (2.1) and let \(\xi_k\) be a centered stochastic sequence satisfying (2.2). Assume that one of the following three conditions is satisfied:

(a) \(\{\xi_k\}\) is \(\varphi\)-mixing.

(b) \(\{\xi_k\}\) is \(\rho\)-mixing and \(\sum k \rho(2^k) < \infty\).

(c) For a certain \(\delta > 0\), \(\{\xi_k\}\) is strongly mixing, \(|\xi_k|^{2+\delta}\) is uniformly integrable, \(\inf_k \text{var} \xi_k > 0\) and \(\sum_n n^{2+\delta}\text{var} \xi_k(n) < \infty\).

Then (2.3) holds.
Remark 2.1. In Theorem 2.2(c) the condition \( \inf_k \text{var} \xi_k > 0 \) can be removed but this requires further additional work and it will be considered elsewhere.

Our last theorem refers to associated sequences of random variables.

Definition 2.4. We call the family \((X_1, \ldots, X_n)\) associated if for any coordinatewise nonincreasing functions \( f(x_1, \ldots, x_n) \) and \( g(x_1, \ldots, x_n) \) we have
\[
\text{cov}(f(X_1, \ldots, X_n), g(X_1, \ldots, X_n)) \geq 0
\]
whenever this is defined.

A sequence \((X_i)\) is called associated if every finite family of variables is associated.

The next theorem extends the well known CLT of Newman and Wright (1981) to general weights while keeping the same conditions as in the classical case.

Theorem 2.3. Assume \( \{a_n\} \) are nonnegative numbers satisfying (2.1). Let \( \{\xi_k\} \) be an associated sequence of random variables which is centered, satisfies (2.2) and
\[
\sum_{j:|k-j|\geq u} \text{cov}(\xi_k, \xi_j) \to 0 \quad \text{as } u \to \infty \text{ uniformly in } k \geq 1.
\]
Then (2.3) holds.

3. Proofs.

3.1. Proof of Theorem 2.1. In order to prove Theorem 2.1 we shall apply Theorem 3.2 from Hall and Heyde (1980) which is stated in the Appendix for convenience (Theorem A). It is easy to see that under (2.1) and (2.2) for every \( \varepsilon > 0 \),
\[
\sum_{i=1}^{n} a_{ni}^2 E \xi_i^2 1(|a_{ni} \xi_i| > \varepsilon) \to 0 \quad \text{as } n \to \infty,
\]
and therefore, for every \( \varepsilon > 0 \),
\[
E \max_{1 \leq i \leq n} a_{ni}^2 \xi_i^2 1(|a_{ni} \xi_i| > \varepsilon) \to 0 \quad \text{as } n \to \infty,
\]
which proves (a) and (c) of Theorem A. In order to verify (b) we shall prove the following lemma which will conclude the proof of Theorem 2.1.

Lemma 3.1. Suppose that \( f \) is a continuous function such that \( \{X_k = f(\xi_k)\} \) is a uniformly integrable family in \( L_1 \). Let \( \{t_{nk}; 1 \leq k \leq n\} \) be a triangular array of real numbers such that
\[
\sup_n \sum_{k=1}^{n} |t_{nk}| \equiv T < \infty
\]
\[\lim_{n \to \infty} \max_{1 \leq k \leq n} |t_{nk}| = 0.\]

Then \(\sum_{k=1}^{n} (X_k - EX_k)t_{nk} \to_p 0\) as \(n \to \infty\).

**Proof.** Let \(M > 1\) and \(0 < \varepsilon < 1\) be two real numbers and define
\[
M_\varepsilon = \lceil M/\varepsilon \rceil + 1, \quad \delta(\varepsilon) = \sup_{|t|, |s| \leq 2M : |t-s| \leq \varepsilon} |f(t) - f(s)|.
\]

For \(1 \leq k \leq n\), denote by \(Y_k^\varepsilon = \sum_{j=-M_\varepsilon}^{M_\varepsilon} f(j\varepsilon)I(\xi_k \leq j\varepsilon < (j+1)\varepsilon)\). From the definition of \(Y_k^\varepsilon\) we deduce that for every \(k \geq 1\),
\[
\left| \left[ X_k (M_\varepsilon \varepsilon \leq \xi_k < M_\varepsilon \varepsilon + \varepsilon) - Y_k^\varepsilon \right] \leq \delta(\varepsilon) \quad \text{a.s.,}
\]
whence, by Jensen inequality and (3.2),
\[
E \left( \left| \sum_{k=1}^{n} (X_k - Y_k^\varepsilon)_t_{nk} \right| - \left( \sum_{k=1}^{n} (EX_k - EY_k^\varepsilon)_t_{nk} \right) \right) \\
\leq 2E \left| \sum_{k=1}^{n} (X_k - Y_k^\varepsilon)_t_{nk} \right| \leq 2T \sup \left| X_k - Y_k^\varepsilon \right| \\
\leq 2T \sup \left| X_k \right| (|\xi_k| > M) + 2T \delta(\varepsilon),
\]
whence the facts that \((X_k)\) is a uniformly integrable family in \(L_1\), \((\xi_k)\) is bounded in probability and \(f\) is continuous show that it is enough to prove the validity of the lemma for \(Y_k^\varepsilon\) with \(M\) and \(\varepsilon\) fixed.

Denote by \(X_k^{\varepsilon,j} = I(\xi_k < j\varepsilon)\). We have
\[
\sum_{k=1}^{n} (Y_k^\varepsilon - EY_k^\varepsilon)_t_{nk} = \sum_{j=-M_\varepsilon}^{M_\varepsilon} f(j\varepsilon) \left[ \sum_{k=1}^{n} (X_k^{\varepsilon,j+1} - EX_k^{\varepsilon,j+1})_t_{nk} \\
- \sum_{k=1}^{n} (X_k^{\varepsilon,j} - EX_k^{\varepsilon,j})_t_{nk} \right]
\]
and therefore it remains to establish the lemma for \(X_k^{\varepsilon,j}\) with \(\varepsilon\) and \(j\) fixed.

Denote by \(\Gamma_i = \sup_j \text{cov}(X_k^{\varepsilon,j}, X_k^{\varepsilon,j+i})\) and by the condition of pairwise mixing
\[\Gamma_i \to 0 \quad \text{as} \quad i \to \infty.\]

With the above notation we have the following estimate:
\[
\text{var} \left( \sum_{k=1}^{n} t_{nk} X_k^{\varepsilon,j} \right) \leq \sum_{k=1}^{n} t_{nk}^2 \text{var}(X_k^{\varepsilon,j}) + 2 \sum_{i=1}^{n-1} \Gamma_i \sum_{k=1}^{n-i} |t_{nk}t_{n,k+i}|.
\]

Obviously we have
\[
\sum_{k=1}^{n} t_{nk}^2 \text{var}(X_k^{\varepsilon,j}) \leq T \max_{1 \leq k \leq n} |t_{nk}|.
\]
To estimate the second term in (3.5) we split the sum in two, one up to \( h \) and another after \( h \), where \( h \) is an integer:

\[
\sum_{i=1}^{n-1} \sum_{k=1}^{n-i} |t_{nk} t_{n,k+i}| \leq h \max_{1 \leq i \leq h} \sum_{k=1}^{n-i} |t_{nk}| |t_{n,k+i}|
\]

(3.7)

\[
+ \max_{h \leq i \leq n} \sum_{k=1}^{n-i} |t_{nk}| |t_{n,k+i}|
\]

\[
\leq hT \max_{1 \leq k \leq n} |t_{nk}| + T^2 \max_{h \leq i \leq n} |\Gamma_i|.
\]

By (3.5), (3.6) and (3.7),

\[
\text{var} \left( \sum_{k=1}^{n} t_{nk} X_k^j \right) \leq T(2h + 1) \max_{1 \leq k \leq n} |t_{nk}| + 2T^2 \max_{h \leq i \leq n} |\Gamma_i|
\]

and the result follows by (3.3) and (3.4) by letting first, \( n \to \infty \), and after, \( h \to \infty \). \( \square \)

3.2. Proof of Corollary 2.1. Without restricting the generality, we can assume \( \sup_k E \xi_k^2 = 1 \). We have

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

In order to apply Theorem 2.1, we fix \( W_n \) such that \( \sum_{|j|>W_n} a_j^2 < n^{-3} \) and take \( k_n = W_n + n \). Then

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

and we have the following estimate

\[
\text{var}(U_n) \leq \sum_{|j|>k_n} \left( \sum_{k=1}^{n} a_{k+j} / \sigma_n \right)^2 \leq n\sigma_n^{-2} \sum_{|j|>k_n} \sum_{k=1}^{n} a_{k+j}^2 \leq n^2\sigma_n^{-2} \sum_{|j|>k_n-n} a_j^2 \leq n^2\sigma_n^{-2} \sum_{|j|>W_n} a_j^2 \leq n^{-1}\sigma_n^{-2} \to 0 \quad \text{as} \quad n \to \infty.
\]

Therefore we have only to prove that \( T_n \to_d N(0,1) \) as \( n \to \infty \). According to Theorem 2.1 it is sufficient to show that

\[
\sup_{-\infty < k < \infty} \left| \frac{\sum_{j=1}^{n} a_{k+j}}{\sigma_n} \right| \to 0 \quad \text{as} \quad n \to \infty.
\]

Let us suppose on the contrary that for some \( \varepsilon > 0 \) there exist a subsequence \((j',n')\), \( n' \to \infty \) such that

\[
\left| \sum_{k=1}^{n'} a_{k+j'} \right| > \varepsilon \sigma_n.
\]
Denote by $A = \sup_{-\infty < k < \infty} |a_k|$ and notice that for $r > j'$,
\[
\left| \sum_{k=1}^{n'} a_{k+r} \right| \geq \varepsilon \sigma_n - 2 A (r - j').
\]
Hence
\[
\frac{\sigma_n^2}{b} \geq \sum_{i=j'}^{j'+W} \left( \sum_{k=1}^{n'} a_{k+i} \right)^2 \geq W \varepsilon^2 \sigma_n^2 - 4 A \varepsilon \sigma_n \left( \sum_{i=j'}^{j'+W} (i - j') \right)
\geq W \varepsilon^2 \sigma_n^2 - 4 A \varepsilon \sigma_n \varepsilon W^2.
\]
Taking $W$ to be the least integer greater than or equal to $3/b\varepsilon^2$ and because $\sigma_n \to \infty$, we obtain for $n'$ sufficiently large,
\[
\frac{\sigma_n^2}{b} \geq \frac{3 \sigma_n^2}{b} - \frac{36 A \varepsilon \sigma_n^3}{b^2 \varepsilon^3} \geq \frac{2 \sigma_n^2}{b},
\]
which is a contradiction. \(\square\)

3.3. Construction of Example 2.1. Let \((Z_i)\) and \((Y_i)\) be two independent sequences of random variables such that \(P(Z_1 = 0) = P(Z_1 = 1) = 1/2\), 
\(Z_{i+1} := 1 - Z_i\) for all \(i \geq 1\) and \((Y_i)\) is an i.i.d. sequence of standard normal variables. Define \(\xi_i := Z_i Y_i\) for \(i \geq 1\). It is easy to verify that \((\xi)\) is a strictly stationary sequence of martingale differences with \(\text{var}(\xi) = 1/2\). We shall prove that \((\xi)\) is ergodic by verifying [cf. Shiryayev (1984), chapter V] that for every measurable bounded function \(f\) and every \(k \geq 1\) positive integer we have
\[
\frac{1}{n} \sum_{i=1}^{n} f(\xi_{i+1}, \ldots, \xi_{i+k}) \to Ef(\xi_1, \ldots, \xi_k) \quad \text{a.s.}
\]
as \(n \to \infty\).

Denote by \(X_i = f(\xi_{i+1}, \ldots, \xi_{i+k})\). If we fix \(Z_1\), say \(Z_1 = 1\), then the sequence \((\xi)\) becomes \(Y_1, 0, Y_3, 0, \ldots\). If \(Z_1 = 0\), the sequence \((\xi)\) consists of \(0, Y_2, 0, Y_4, \ldots\).

Therefore for \(Z_1\) fixed, the sequence \((X_i)\) becomes stationary and \(k\)-dependent, therefore ergodic, and we have
\[
\begin{align*}
I(Z_1 = 1) & \frac{1}{n} \sum_{i=1, i: \text{even}}^{n} X_i \to I(Z_1 = 1) \frac{1}{2} Ef(Y_1, 0, Y_3, \ldots) \quad \text{a.s.}, \\
I(Z_1 = 1) & \frac{1}{n} \sum_{i=1, i: \text{odd}}^{n} X_i \to I(Z_1 = 1) \frac{1}{2} Ef(0, Y_3, 0, \ldots) \quad \text{a.s.}, \\
I(Z_1 = 0) & \frac{1}{n} \sum_{i=1, i: \text{even}}^{n} X_i \to I(Z_1 = 0) \frac{1}{2} Ef(0, Y_2, 0, \ldots) \quad \text{a.s.}, \\
I(Z_1 = 0) & \frac{1}{n} \sum_{i=1, i: \text{odd}}^{n} X_i \to I(Z_1 = 0) \frac{1}{2} Ef(Y_2, 0, Y_4, \ldots) \quad \text{a.s.}.
\end{align*}
\]
Now we add all these relations, and by stationarity and construction we have:

\[
\frac{1}{n} \sum_{i=1}^{n} X_i \rightarrow \frac{1}{2} \left[ \text{Ef}(Y_1, 0, Y_3, \ldots) + \text{Ef}(0, Y_2, 0, Y_4, \ldots) \right]
\]

\[= \text{Ef}(X_1) \quad \text{a.s.}\]

Let us now consider the numbers \((a_{ni})\) satisfying (2.1), (2.2) and in addition \(\sum_{i=1}^{n} a_{ni}^2 = 1/3\) for all \(n \geq 2\). Set \(S_n := \sum_{i=1}^{n} a_{ni} \xi_i\) and let \(Y\) be a standard normal random variable independent of \(Z_i\). By the construction \(S_n\) has the same distribution as \(Y \sqrt{\sum_{i=1}^{n} a_{ni}^2} Z_i\). Because \(Z_i = Z_i\) for every \(i \geq 1\) we have

\[\sum_{i=1}^{n} a_{ni}^2 Z_i = Z_1 + \sum_{i=1, i: \text{odd}}^{n} a_{ni}^2 + Z_2 + \sum_{i=1, i: \text{even}}^{n} a_{ni}^2 = 1/3 Z_1 + 5/3 Z_2 = 5/3 - 4/3 Z_1.\]

Therefore \(S_n\) has a fixed nongaussian distribution for all \(n \geq 2\).

3.4. Proof of Theorem 2.2(a), (b). In order to prove Theorem 2.3 under the assumptions (a) and (b), according to Theorem 2.1 and Theorem 4.1, respectively, in Utev (1990) we have only to verify the Lindeberg's condition

\[\sum_{i=1}^{n} \text{Ea}_{ni}^2 \xi_i^2 1(|a_{ni} \xi_i| > \varepsilon) \rightarrow 0 \quad \text{as} \quad n \rightarrow \infty \quad \text{for every} \quad \varepsilon > 0\]

and this follows exactly as (3.1) by (2.1) and (2.2). □

3.5. Proof of Theorem 2.2(c). In order to prove this part of Theorem 2.2, we shall first use a truncation argument and after that we shall apply Theorem B from the Appendix. The proof requires the following auxiliary lemma.

**Lemma 3.2.** Assume \((X_i)\) satisfies the conditions of Theorem 2.2(c) then

\[\text{var} \left( \sum_{i=a}^{b} a_{ni} X_i \right) \leq C \sum_{i=a}^{b} a_{ni}^2\]

where \(C = \sup_k \text{E} X_k^2 + c(\delta)(\sum_i n^{2/3}(n))^{\delta/(2+\delta)} \sup_k \|X_k\|_{2+\delta}^2\), and \(c(\delta)\) is a numerical constant depending on \(\delta\).

**Proof.** We have

\[\text{var} \left( \sum_{i=a}^{b} a_{ni} X_i \right) \leq \sum_{i=a}^{b} a_{ni}^2 \text{var} X_i + 2 \sum_{j=a}^{b-1} \sum_{i=j+1}^{b} a_{ni} a_{nj} \text{cov}(X_j, X_i)\]

\[\leq \sum_{i=a}^{b} a_{ni}^2 \text{var} X_i + 2 \sum_{i=a}^{b} a_{ni}^2 \sum_{j=a, j \neq i}^{b} |\text{cov}(X_i, X_j)|\]

whence the result follows by Lemma B from the Appendix.
In order to prove Theorem 2.2(c), we shall truncate first the variables at the level \( A > 0 \) and denote
\[
\xi_i' = \xi_i I(|\xi_i| \leq A) - E\xi_i I(|\xi_i| \leq A)
\]
and
\[
\xi_i'' = \xi_i I(|\xi_i| > A) - E\xi_i I(|\xi_i| > A).
\]
By Lemma 3.2 there is a positive constant \( C \) such that
\[
\text{var}(\sum_{i=1}^{n} a_{ni} \xi_i'') \leq C \left[ \sup_k (E\xi_k^2 I(|\xi_k| > A)) + \sup_k (E|\xi_k|^{2+\delta} I(|\xi_k| > A))^{2/(2+\delta)} \right]
\]
and notice that by the uniform integrability of \(|\xi_k|^{2+\delta}\)
\begin{equation}
\lim_{A \to \infty} \sup_n \text{var} \left( \sum_{i=1}^{n} a_{ni} \xi_i'' \right) = 0.
\end{equation}
By (2.2) and (3.9) we obtain
\begin{equation}
\lim_{A \to \infty} \text{var} \left( \sum_{i=1}^{n} a_{ni} \xi_i' \right) = 0 \quad \text{uniformly in } n.
\end{equation}
By (3.9), (3.10) and Theorem 4.2 in Billingsley (1968), in order to prove the theorem it is enough to show that for every fixed positive \( A \) we have
\[
\frac{\sum_{i=1}^{n} a_{ni} \xi_i'}{\text{std \, dev}(\sum_{i=1}^{n} a_{ni} \xi_i')} \to_N (0, 1) \quad \text{as } n \to \infty.
\]
To prove this we shall verify the conditions of Theorem B given in the Appendix. The condition (a) requires a uniform bound on the variance of partial sums which follows by Lemma 3.2. The Lindeberg condition (d) is satisfied by (2.1), (2.2) and (3.10). Condition (b) is a consequence of (2.1), (2.2) and (3.10). We have only to verify condition (c). Let \( \varepsilon = 1/(2 + 2\delta) \).

Notice that by Lemma A in the Appendix, we can find a constant \( K_1 \) such that
\begin{equation}
\left| \text{cov} \left( \left[ \exp \left( \text{i} t \sum_{j=a}^{b} a_{nj} \xi_j' \right) \right], \left[ \exp \left( \text{i} \sum_{k=b+u}^{c} a_{nk} \xi_k' \right) \right] \right) \right|
\end{equation}
\begin{equation}
= \left| \text{cov} \left( \left[ \exp \left( \text{i} \sum_{j=a}^{b} a_{nj} \xi_j' \right) - 1 \right], \left[ \exp \left( \text{i} \sum_{k=b+u}^{c} a_{nk} \xi_k' \right) - 1 \right] \right) \right|
\end{equation}
\begin{equation}
\leq K_1 t^{2\alpha/(1+\delta)}/(u) \left\| \sum_{j=a}^{b} a_{nj} \xi_j' \right\|_{2+2\delta} \left\| \sum_{j=b+u}^{c} a_{nj} \xi_j' \right\|_{2+2\delta}.
\end{equation}
Notice now that by Lemma 3.2 and the Hölder inequality, we find a constant $K_2$ such that

$$
\mathbb{E}\left[\sum_{i=a}^{b} a_{ni} \xi_i^\delta\right]^{2+2\delta} \leq (2A)^{2\delta} \mathbb{E}\left(\sum_{i=a}^{b} a_{ni}^\delta\right)^2 \left(\sum_{i=a}^{b} |a_{ni}|\right)^{2\delta}
$$

$$
\leq K_2 (2A)^{2\delta} \left(\sum_{i=a}^{b} a_{ni}^2\right) \left(\sum_{i=a}^{b} a_{ni}^2\right)^{\delta} (b - a + 1)^\delta
$$

$$
= K_2 (2A)^{2\delta} \left(\sum_{i=a}^{b} a_{ni}^2\right)^{(1+\delta)} (b - a + 1)^\delta.
$$

Therefore the right-hand side of (3.11) is bounded above by

$$
K_3 t^2 \left[\alpha(u)(c - a)\right]^{\delta/(1+\delta)} \sum_{i=a}^{b} a_{ni}^2,
$$

where $u = (c - a)^{1-\varepsilon}$ and $K_3$ is a constant.

Due to our condition on $\alpha(n)$ and the selection of $\varepsilon$, we have

$$
\sum_i (\alpha(2^i) 2^{i/(1-\varepsilon)})^{\delta/(1+\delta)} \leq K_4 \sum_i \left(2^i\right)^{-(1+\delta)/\delta} 2^{i/(1-\varepsilon)} \delta/(1+\delta)
$$

$$
\leq K_5 \sum_i (2^{1/(1+2\delta)})^{-i} < \infty,
$$

where $K_4$ and $K_5$ are constants, which completes the proof of Theorem 2.2 under (c). □

### 3.6. Proof of Theorem 2.3.

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,

$$
0 \leq \left(\sum_{i=a}^{b-u} a_{ni} \sum_{j=i+u}^{b} a_{nj} \text{cov}(\xi_k, \xi_l)\right)
$$

$$
\leq \sup_k \left(\sum_{j:|k-j|\geq u} \text{cov}(\xi_k, \xi_l)\right) \left(\sum_{i=a}^{b} a_{ni}^2\right).
$$

(3.12)

In particular by (2.4), there exists a constant $C$ such that for every $1 \leq a \leq b \leq n$,

$$
\text{var}\left(\sum_{i=a}^{b} a_{ni} \xi_i\right) \leq C \sum_{i=a}^{b} a_{ni}^2.
$$

We shall construct now a triangular array of random variables $(Z_{ni}, 1 \leq i \leq n)$ for which we shall make use of Theorem C in the Appendix. Fix a small positive $\varepsilon$ and find a positive integer $u = u_\varepsilon$ such that

$$
0 \leq \left(\sum_{i=1}^{n-u} a_{ni} \sum_{j=i+u}^{b} a_{nj} \text{cov}(\xi_i, \xi_j)\right) \leq \varepsilon \quad \text{for every } n \geq u + 1.
$$
This is possible because of (3.12) and (2.4). Denote by \( \lfloor x \rfloor \) the integer part of \( x \) and define

\[
K = \left\lfloor \frac{1}{\varepsilon} \right\rfloor.
\]

\[
Y_{nj} = \sum_{i=uj+1}^{uj+2} a_{ni} \xi_i, \quad j = 0, 1, \ldots,
\]

\[
A_j = \left\{ i \colon 2K_j \leq i < 2K_j + K, \, \text{cov}(Y_{nj}, Y_{n,i+1}) \leq \frac{2}{K} \sum_{i=2K_j}^{2K_j+K} \text{var}(Y_{ni}) \right\}.
\]

Since \( 2 \text{cov}(Y_{nj}, Y_{n,i+1}) \leq \text{var}(Y_{nj}) + \text{var}(Y_{n,i+1}) \), we get that for every \( j \) the set \( A_j \) is not empty. Now we define the integers \( m_1, m_2, \ldots, m_n \) recurrently by \( m_0 = 0 \):

\[
m_{j+1} = \min\{m ; m > m_j, m \in A_j\}
\]

and define

\[
Z_{nj} = \sum_{i=m_j+1}^{m_{j+1}} Y_{ni}, \quad j = 0, 1, \ldots,
\]

\[
\Delta_j = \{u(m_j + 1) + 1, \ldots, u(m_{j+1} + 1)\}.
\]

We observe that

\[
Z_{nj} = \sum_{k \in \Delta_j} a_{nk} X_k, \quad j = 0, 1, \ldots.
\]

It is easy to see that every set \( \Delta_j \) contains no more than \( 3Ku \) elements. Hence, for every fixed positive \( \varepsilon \) by (2.1) and (2.2) the array \( (Z_{nj} : i = 1, \ldots, n; \, n \geq 1) \) satisfies the Lindeberg's condition. It remains to observe that by Theorem C and construction

\[
\left| \mathbb{E} \exp \left( i \sum_{j=1}^{n} Z_{nj} \right) - \prod_{j=1}^{n} \mathbb{E} \exp(iZ_{nj}) \right| 
\]

\[
\leq c \varepsilon^2 \left( \text{var}\left( \sum_{j=1}^{n} Z_{nj} \right) - \sum_{j=1}^{n} \text{var}(Z_{nj}) \right)
\]

\[
\leq c \varepsilon^2 \left( 2 \left( \sum_{i=1}^{n} \text{cov}(Z_{ni}, Z_{n,i+1}) + 2 \sum_{i=1}^{n-2} \sum_{j=i+2}^{n} \text{cov}(Z_{ni}, Z_{nj}) \right) \right)
\]

\[
\leq c \varepsilon^2 \left( 4 \sum_{i=1}^{n-u} a_{ni} \sum_{j=i+u}^{n} a_{nj} \text{cov}(\xi_i, \xi_j) + 2 \sum_{j=1}^{n} \text{cov}(Y_{nmj}, Y_{n,mj+1}) \right)
\]
Now the proof is complete by Theorem 4.2 in Billingsley (1968). □

APPENDIX

This section contains some of the theorems which were used in the proofs of the results of Section 3.

The following lemma is a variant of Theorem 17.2.2 in Ibragimov and Linnik (1971). The constant comes from Theorema 1.1 in Bradley and Bryc (1985).

**Lemma A.** Let $\xi, \eta$ be two complex-valued random variables measurable with respect to $\mathcal{A} = \sigma(\xi)$ and $\mathcal{B} = \sigma(\eta)$. Let $\delta > 0$ and assume $E|\xi|^{2+\delta} < \infty$ and $E|\eta|^{2+\delta} < \infty$; then

$$|E(\xi \eta) - E(\xi) E(\eta)| \leq 2\pi \alpha^{\delta/(2+\delta)}(\mathcal{A}, \mathcal{B}) \|\xi\|_{2+\delta}\|\eta\|_{2+\delta}.$$

By the application of Theorem 1.2 in Rio (1993), followed by the Cauchy–Schwarz inequality, we formulate the following result.

**Lemma B.** Let $(X_n)$ be a strongly mixing sequence of random variables such that $E|X_k|^{2+\delta} < \infty$ for a certain $\delta > 0$ and every $k \geq 1$. Then there is a numerical constant $c(\delta)$ depending only on $\delta$ such that for every $k \geq 1$ we have

$$\sum_{n=k}^{k+m} |\text{cov}(X_k, X_n)| \leq c(\delta) \left( \sum_{n=1}^{m} n^{2/3} \alpha(n) \right)^{\delta/(2+\delta)} \sup_{j} \|X_j\|_{2+\delta}.$$

The following theorem is a simplified version of Theorem 3.2 from Hall and Heyde (1980).

**Theorem A.** Let $(X_{ni})$, $1 \leq i \leq k_n$, be a square integrable martingale difference sequence under the natural filtrations.

Suppose that:

(a) $\max_{1 \leq i \leq k_n} |X_{ni}| \to_{P} 0$;
(b) $\sum_{i=1}^{k_n} X_{ni}^2 \to_{P} 1$;
(c) $\text{E} \max_{1 \leq i \leq k_n} X_{ni}^2$ is bounded in $n$.

Then $S_n \to_{D} N(0, 1)$ as $n \to \infty$ where $S_n = \sum_{i=1}^{k_n} X_{ni}$. 
A careful analysis of the proof of Theorem 4.1 in Utev (1990) gives the following statement.

**THEOREM B.** Let \( \{X_{ni}, 1 \leq i \leq k_n\} \) be a triangular array of random variables such that the following hold.

(a) \( \text{var}(\sum_{j=a}^{b} X_{nj}) \leq C \sum_{j=a}^{b} \text{var}(X_{nj}) \) for every \( 0 \leq a \leq b \leq k_n \) where \( C \) is a universal constant;

(b) \( \liminf_{n \to \infty} \frac{\text{var}(\sum_{j=1}^{k_n} X_{nj})}{\sum_{j=1}^{k_n} \text{var}(X_{nj})} > 0; \)

(c) \( \left| \text{cov}\left( \exp\left( it \sum_{j=a}^{b} X_{nj}\right), \exp\left( it \sum_{j=b+u}^{c} X_{nj}\right) \right) \right| \leq h(u) \sum_{j=a}^{c} \text{var}(X_{nj}) \)

for every \( 1 \leq a \leq b \leq c \leq k_n \) where \( h(u) \geq 0 \), \( \sum h_i(\varepsilon^2) < \infty \) and \( u \) is of the form \( u = \lfloor (c - a)^{1-\varepsilon} \rfloor \) for a certain \( 0 < \varepsilon < 1; \)

(d) \( \sigma_n^{-2} \sum_{i=1}^{k_n} \mathbb{E} |X_{ni}|^2 \mathbb{I}(\{|X_{ni}| > \varepsilon \sigma_n\}) \to 0 \) as \( n \to \infty \) for every \( \varepsilon > 0 \), where \( \sigma_n^{-2} \) denotes \( \text{var}(\sum_{i=1}^{k_n} X_{ni}) \). Then \( S_n/\sigma_n \to N(0, 1) \) as \( n \to \infty \).

The next theorem is taken from Newman and Wright (1981).

**THEOREM C.** Let \( \{Z_i, 1 \leq i \leq k\} \) be an associated family of random variables. Then

\[
\left| \mathbb{E} \exp\left( it \sum_{j=1}^{n} Z_j \right) - \prod_{j=1}^{n} \mathbb{E} \exp(itZ_j) \right| \leq ct^2 \left( \text{var}\left( \sum_{j=1}^{n} Z_j \right) - \sum_{j=1}^{n} \text{var}(Z_j) \right).
\]

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