PROBABILITIES OF MODERATE DEVIATIONS FOR
SOME STATIONARY $\phi$-MIXING PROCESSES

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Probabilities of moderate deviations of the sample mean from the
population mean are calculated for certain strictly stationary $\phi$-mixing
processes.

1. Introduction. Consider a strictly stationary sequence $\{X_n, n \geq 1\}$ of ran-
dom variables (rv's) defined on a probability space $(\Omega, \mathcal{F}, P)$. Define $M^\sigma_{k}$ as
the $\sigma$-field generated by $X_1, \ldots, X_n$ and $M^\sigma_{n+u}$ as the $\sigma$-field generated by
$X_{n+u}, \ldots, X_{n+1}$. It is assumed that

$$P(B|A) = 0 \quad \text{and} \quad B \in M^\sigma_{k+n},$$

(1.1) \quad $$P(B|A) = 0 \quad \text{and} \quad B \in M^\sigma_{k+n},$$

$$|P(B|A) - P(B)| \leq \phi(n),$$

where $\phi(i)$'s are nonnegative numbers satisfying

$$1 \geq \phi(1) \geq \phi(2) \geq \ldots, \lim_{n \to \infty} \phi(n) = 0.$$

The condition (1.1) is usually referred to in the literature as the $\phi$-mixing con-
dition.

Central limit theorems (CLT's) for sequences of rv's satisfying (1.1) are
proved by Ibragimov (1962) under certain conditions on the moments of the
$X_i$'s and on the $\phi(i)$'s. These results ensure under certain conditions convergence
in law of the sample sum $S_n = \sum_{i=1}^n X_i$ (when suitably normalized) to the normal
$(0, 1)$ variable. Berry–Esseen type results for such sums are available in Reznik
(1968) and Philipp (1971), and have been used by them in proving laws of the
iterated logarithm.

In this paper, we obtain probabilities of moderate deviations (PMD) for sample
means. We shall observe that our results generalize the PMD findings of Ghosh
(1974) under $m$-dependence. Included also are the PMD results of Rubin and
Sethuraman (R–S) (1965) in the i.i.d. case. R–S, however, consider also the
case when the $X_i$'s are independent but not identically distributed.

We adopt the convention of denoting by $K_i (\geq 0), K_i (\geq 0), \gamma (\geq 0)$, etc.,
generic constants. The word “strictly” will henceforth be omitted before “sta-
tionary” for brevity. Also, in what follows, $a_n \sim b_n$ will mean $a_n/b_n \to 1$ as
$n \to \infty$.

Received November 7, 1973; revised August 23, 1976.

1 Part of the work was done while the author was visiting the Department of Mathematics,
University of Illinois.

AMS 1970 subject classifications. 60F99, 60G10.

Key words and phrases. Probabilities of moderate deviations, stationary processes, $\phi$-mixing,
Esseen's lemma, $m$-dependent processes.
The main result of the paper is as follows.

**Theorem.** Suppose the stationary process \( \{X_n, n \geq 1\} \) satisfies (1.1). Then, for any \( c > 0 \),

\[
P(n^{-1}S_n - \mu > c\sigma (\log n/n)^{\delta}) \sim (2\pi c^3 \log n)^{-1} n^{-1/2},
\]

and

\[
P(|n^{-1}S_n - \mu| > c\sigma (\log n/n)^{\delta}) \sim 2(2\pi c^3 \log n)^{-1} n^{-1/2}
\]

hold under the conditions

\[
E|X_1|^2^{\delta+\delta} < \infty, \quad \text{for some} \quad \delta > 0;
\]

(1.5) \[
\sum_{n=1}^{\infty} \phi(n) < \infty;
\]

(1.6) \[
\sigma^2 = V(X_1) + 2 \sum_{i=1}^{\infty} \text{Cov}(X_i, X_{i+1}).
\]

The proof of the theorem is postponed to Section 3. Certain basic lemmas pertaining to the proof of the theorem are proved in Section 2. The results (1.2) and (1.3) are the PMD results. Noting that \( x(1 - \Phi(x))/N(x) \to 1 \) as \( x \to \infty \), where \( N(x) = (2\pi)^{-1/2} \exp(-1/2x^2), \Phi(x) = \int_{-\infty}^{x} N(y) \, dy \), alternate representations of (1.2) and (1.3) are as follows:

\[
P(n^{-1}(S_n - n\mu)/\sigma > c(\log n)^{\delta}) \sim 1 - \Phi(c(\log n)^{\delta});
\]

(1.7) \[
P(n^{-1}|S_n - n\mu|/\sigma > c(\log n)^{\delta}) \sim 2[1 - \Phi(c(\log n)^{\delta})].
\]

The calculations of Ibragimov (1962) show that \( \sigma^2 < \infty \) under our conditions. Our conditions (1.4) and (1.5) are stronger than the ones required in proving the corresponding CLT's. Ibragimov (1962) has proved a CLT (see his Theorem 1.5) assuming (1.5) and the weaker condition \( E|X_1|^2 < \infty \). For proving a Berry–Esseen type result Reznik (1968) needs the stronger assumption \( E|X_1|^{2+\delta} < \infty \) for some \( \delta > 0 \) in place of \( E|X_1|^2 < \infty \). It is not out of place to mention here that Reznik's (1968) paper contains a gap. He applies Esseen's lemma to two distribution functions, \( F_1 \) and \( F_2 \), without verifying that at least one of them satisfies some smoothness condition (see Feller (1968), page 512, Lemma 2). The gap has been closed by Stout (1974).

The moment condition (1.4) is assumed by R–S (1965) and Ghosh (1974). If \( \phi(n) = 0 \) for \( n \geq m + 1 \), where \( m \) is some fixed nonnegative integer, we obtain the case considered by Ghosh (1974). Thus the present theorem includes the \( m \)-dependent as well as the i.i.d. cases earlier considered.

2. Some basic lemmas. We develop here some lemmas to be used subsequently in proving the theorem. From now on, it is assumed without any loss of generality that \( \mu = 0 \) and \( \sigma = 1 \). Define

\[
X_i' = X_{n,i} = X_i \quad \text{if} \quad |X_i| \leq n^{1/2},
\]

\[
= 0 \quad \text{otherwise},
\]

\( 1 \leq i \leq n, \ n \geq 1 \).

Then, if \( S'_n = \sum_{i=1}^{n} X_i' (n \geq 1) \), one gets the following lemma.
LEMMA 1. Under (1.4),
\[(2.2) \quad |P(S_n > c(n \log n)^{1/2}) - P(S_n > c(n \log n)^{1/2})| = O(-n^{-1/2}).\]

PROOF.
LHS of (2.2) \[\leq P\left(\bigcup_{i=1}^{n} \{X_i > n^{1/2}\}\right) \leq nP\{|X_i > n^{1/2}\}\]
\[\leq nn^{-1(2x_2+\delta)} E|X_i|^{2x_2+\delta} = O(n^{-1(2x_2+\delta)}),\]
using the definition of $X_i$'s, the stationarity of the $X_i$'s, Markov's inequality and (1.4). This proves the lemma.

In view of the above lemma, to prove (1.2), it suffices to show that
\[(2.3) \quad P(S_n > c(n \log n)^{1/2}) \sim (2\pi c^2 \log n)^{-1/2}.\]
We prove one more lemma in this section which gives the order of $E|S_n|^m$ for all positive integers $m \geq 2$. The present lemma generalizes Lemma 1.9 of Ibragimov (1962), but only for the truncated random variables $X_i$'s.

LEMMA 2. For $1 \leq u \leq n$, and for any positive integer $m \geq 2$,
\[(2.4) \quad E|S_n|^m \leq K_1(uR(m)(\log n)^{x_2} + u^{1/m}),\]
where $R(m) = R_0(m) = n^{1/(m-\delta_2-\delta)}$ or 0 according as $m > 0 \leq 2 + \delta$; $K_1$ and $K_2$ are constants which may depend on $m$ but not on $n$ and $u$.

PROOF. We prove the lemma by induction. First observe that
\[(2.5) \quad E|S_n|^2 = uE(X_i^2) + 2 \sum_{j=1}^{n} \sum_{i=j}^{n} E(X_i^2X_{i+j}^2),\]
using the stationarity of the $X_i$'s. Using now a result of Ibragimov (1962), (see, e.g., Lemma 1, page 170 of Billingsley (1968)), one gets
\[(2.6) \quad |E(X_i^2X_{i+j}^2)| \leq |E(X_i^2X_{i+j}^2) - E(X_i^2)E(X_{i+j}^2)| + |E(X_i^2)E(X_{i+j}^2)|
\[\leq 2\phi^2(j)E(X_i^2) + |EX_i|^2.\]
Now, from the definition of the $X_i$'s in (2.1), the stationarity of the $X_i$'s, Markov's inequality, and (1.4) one gets,
\[(2.7) \quad |EX_i| \leq E[|X_i|I_{|X_i|>x_2}] \leq n^{-1(2x_2+\delta)}E[|X_i|^{2x_2+\delta}I_{|X_i|>x_2}] \leq o(n^{-1(2x_2+\delta)}),\]
where $I$ is the usual indicator function. Also, $EX_i^2 \leq EX_i^2 = A$ (say) $< \infty$. It follows now from (2.5)—(2.7) that
\[(2.8) \quad E|S_n|^2 \leq Au + 2 \sum_{j=1}^{n} \sum_{i=j}^{n} [2\phi^2(j) + o(n^{-1(2x_2+\delta)})] \leq K_1 u,
\]
where one uses $u \leq n$ and (1.5). This proves the lemma for $m = 2$.

To prove the lemma by induction, assume next that
\[(2.9) \quad C(u, m) = E|S_n|^m \leq K_1(uR(m)(\log n)^{x_2} + u^{1/m}) = D(u, m) \quad \text{(say)},\]
and prove a similar inequality for $C(u, m + 1)$. Towards this end, define $S'_{u,t} = \sum_{i=1}^{u} X_{u+t+i}$ and $S'_{u,t} = \sum_{i=1}^{u} X_{u+i}$. Using the definition of $C(u, m)$ and the stationarity of the $X_i$'s, one gets
\[(2.10) \quad E(|S_n' + S'_{u,t}|)^{m+1} \leq E(|S_n'| + |S'_{u,t}|)^{m+1} = \sum_{j=0}^{m+1} \binom{m+1}{j} E(|S_n'|^{m+1-j}|S'_{u,t}|^j)
\[= 2C(u, m + 1) + \sum_{j=1}^{m} \binom{m+1}{j} E(|S_n'|^{m+1-j}|S'_{u,t}|^j).\]
Using the stationarity of the \( X_i \)'s, and Lemma 1, page 170 of Billingsley (1968) with \( r = (m + 1)/(m + 1 - j) \), and \( s = (m + 1)/j \), one gets
\[
|E|S'_u|^{m + 1 - j}|S'_{u,t}|^{j}| - E|S'_u|^{m + 1 - j}|E|S'_{u,t}|^{j}|
\leq 2\phi^{(m + 1 - j)/(m + 1 - j)}(1)(E|S'_u|^{m + 1 - j}/(m + 1))/(E|S'_u|^{m + 1 - j}/(m + 1))^{j/(m + 1)}
\leq 2\phi^{j/(m + 1)}(t)C(u, m + 1), \quad 1 \leq j \leq m.
\]

It follows now from (2.10) and (2.11) that
\[
E|S'_u + S'_{u,t}|^{m + 1} \leq (2 + 2\phi^{j/(m + 1)}(t)2^{m + 1})C(u, m + 1)
+ \sum_{j=1}^{m} (m/j) E|S'_u|^{m + 1 - j}E|S'_{u,t}|^{j}
\leq (2 + 2^{m+2}\phi^{j/(m + 1)}(t))C(u, m + 1)
+ \sum_{j=1}^{m} (m/j)(E|S'_u|^{m + 1 - j}/(m + 1))(E|S'_{u,t}|^{j}/m)
\leq (2 + 2^{m+2}\phi^{j/(m + 1)}(t))C(u, m + 1)
+ 2^{m+1}C(M^{m+1}/m(u, m)).
\]

Again from (2.9), \( C^{(m+1)/m(u,m)} \leq K_1(u^{(m+1)/m}R^{(m+1)/m}(m)(\log n)K^2 + u^{1/(m+1)}) \). Now, for \( m + 1 \leq c^2 + 2 + \delta \), \( R(m) = R(m + 1) = 0 \). For \( m + 1 > c^2 + 2 + \delta \), \( (uR(m))^{(m+1)/m(u,m)} \leq uK^{1/(m+1)/m(u,m)} = uR(m + 1) \). It follows now from (2.12) that
\[
E|S'_u + S'_{u,t}|^{m + 1} \leq [2 + 2^{m+2}\phi^{j/(m + 1)}(t)]C(u, m + 1) + D(u, m + 1).
\]

Using the Minkowski inequality, one gets
\[
C(2u, m + 1) = E|S'_u + S'_{u,t} + S''_{u,t} - S''_{u,t}|^{m + 1}
\leq \left\{ E^{(m+1)/m}|S'_u + S'_{u,t}|^{m + 1}
+ \left( \sum_{u=1}^{m} + \sum_{m+1}^{m+1} \right) E^{1/(m + 1)}|X'_i|^{m + 1} \right\}^{m + 1}.
\]

Note now that for \( m + 1 \leq c^2 + 2 + \delta \),
\[
E|X'_i|^{m + 1} \leq E|X'_i|^{c^2 + 2 + \delta} \leq E|X'_i|^{c^2 + 2 + \delta} \leq K_1,
\]
and for \( m + 1 > c^2 + 2 + \delta \),
\[
E|X'_i|^{m + 1} \leq n^{1/(m+1)/(m+1-2-2-\delta)}E|X'_i|^{c^2 + 2 + \delta} \leq K_1R(m + 1),
\]
using (1.4), (2.1) and the definition of \( R(m) \). The above leads to
\[
E|X'_i|^{m + 1} \leq K_1(1 + R(m + 1)).
\]

Combining (2.13) and (2.15), one gets from (2.14),
\[
C(2u, m + 1) \leq \left\{ [(2 + 2^{m+2}\phi^{j/(m + 1)}(t))C(u, m + 1) + D(u, m + 1)]^{1/(m + 1)}
+ K_1(1 + R(m + 1))^{1/(m + 1)} \right\}^{m + 1}.
\]

Consider now the two separate cases (i) \( m + 1 \leq c^2 + 2 + \delta \) and (ii) \( m + 1 > c^2 + 2 + \delta \).

**Case (i).** Here \( R(m + 1) = 0 \). Given any \( \varepsilon_1 > 0 \), choose \( t \) to be sufficiently large (not dependent on \( n \)) that \( 2^{m+2}\phi^{j/(m + 1)}(t) < \varepsilon_1 \) (by 1.5)). Then, one gets
from (2.9) and (2.16),

\[ C(2u, m + 1) \leq \left[ \left( 2 + \varepsilon_i \right) C(u, m + 1) + K_i u^{t\left(m+1\right)} \right]^{1/(m+1)} + K_i \]

\[ \leq \left[ \left( 1 + \varepsilon_i \right) \left( 2 + \varepsilon_i \right) C(u, m + 1) + K_i u^{t\left(m+1\right)} \right]^{1/(m+1)} \]

for \( u \) sufficiently large, say \( u \geq u_0 = u_0(m, \varepsilon_i) \). Also, given any \( \varepsilon > 0 \), choose \( \varepsilon_i (> 0) \) such that \( (1 + \varepsilon_i)^{m+1}(2 + \varepsilon_i) \leq 2 + \varepsilon \). This leads to

\[ C(2u, m + 1) \leq \left( 2 + \varepsilon \right) C(u, m + 1) + K_i u^{t\left(m+1\right)}, \]

for \( u \geq u_0 \). This is also valid for \( u \leq u_0 \) by using the inequality \( E[S_i | X_i']^m \leq u^m E[X_i | X_i']^{d^2 + 2} \) and by suitable readjustment of \( K_i \) if necessary. Repeating (2.17) \( r \) times and using \( C(1, m + 1) \leq E[X_i | X_i']^{d^2 + 2} \leq K_i \) one gets

\[ C(2^r, m + 1) \leq \left( 2 + \varepsilon \right)^r C(1, m + 1) \]

\[ + K_i \left( 2^r/2 \right)^{1/(m+1)} \left( 1 + (2 + \varepsilon)2^{-1/(m+1)} + \cdots \right) \]

\[ + \left( 2 + \varepsilon \right)^{-1/(m+1)} \]

\[ \leq K_i \left[ (2 + \varepsilon)^r + (2^r)^{1/(m+1)} \right] \leq K_i (2^r)^{1/(m+1)} = D(2^r, m + 1), \]

choosing \( \varepsilon (> 0) \) so small that \( 2 + \varepsilon < 2^r \leq 2^{1/(m+1)} \) for \( m \geq 2 \).

Case (ii). Here \( R(m + 1) \neq 0 \). Take \( t = \lfloor (\log n)^{m+1/2} \rfloor \), the integer part of \( (\log n)^{m+1/2} \). Then (2.16) and (1.5) lead to

\[ C(2u, m + 1) \]

\[ \leq \left[ \left( 2 + K_i (\log n)^{-1} \right) C(u, m + 1) + D(u, m + 1) \right]^{1/(m+1)} \]

\[ + K_i (\log n)^{1/2} \left[ (2 + K_i (\log n)^{-1}) C(u, m + 1) + D(u, m + 1) \right]^{1/(m+1)} \]

\[ \leq \left[ \left( 1 + K_i (\log n)^{-1} \right) C(u, m + 1) + D(u, m + 1) \right]^{1/(m+1)} \]

(2.19)

\[ \leq \left( 2 + K_i (\log n)^{-1} \right) C(u, m + 1) + D(u, m + 1) \]

\[ \leq 2(1 + K_i (\log n)^{-1}) C(u, m + 1) + K_i (uR(m + 1) (\log n)^{K_1} + u^{1/(m+1)}). \]

When \( 2^r \leq n \), i.e., \( r \leq (\log 2)^{-1}(\log n) \), repeating the inequality (2.19) \( r \) times and recalling \( C(1, m + 1) = E[X_i']^{m+1} \leq K_i R(m + 1) \), one gets

\[ C(2^r, m + 1) \leq 2^r (1 + K_i (\log n)^{-1}) C(1, m + 1) \]

\[ + K_i 2^r R(m + 1) (\log n)^{K_1} \sum_{j=1}^{r-1} \left[ (2 + K_i (\log n)^{-1}) 2^{-j/(m+1)} \right] \]

\[ + K_i (2^r)^{1/(m+1)} \sum_{j=1}^{r-1} \left[ (2 + K_i (\log n)^{-1}) 2^{-j/(m+1)} \right] \]

\[ \leq K_i 2^r R(m + 1) + K_i 2^r R(m + 1) (\log n)^{K_1} + K_i (2^r)^{1/(m+1)} \]

\[ \leq D(2^r, m + 1). \]

Combining (2.18) and (2.20) one gets for any \( u = 2^r (\leq n) \),

\[ C(2^r, m) \leq K_i [2^r R(m) (\log n)^{K_1} + (2^r)^{1/m}]. \]

Now suppose \( 2^r < u < 2^{r+1} \). Use the binary decomposition \( u = \sum_{j=0}^{r-1} \nu_j 2^j \) with \( \nu_j = 1 \), \( \nu_j = 0 \) or 1 for \( 0 \leq j \leq r - 1 \). Then,

\[ E[\sum_{i=1}^{n} X_i']^{m} = E[(\sum_{i=1}^{n} \nu_j 2^j + \cdots + \nu_j 2^{r-1})X_i']^{m}, \]

(2.22)
where \( i = \nu_{r+1-j} 2^{r+1-j} (1 \leq j \leq r + 1) \), \( \sum_{m=1}^{\infty} X_j \) is interpreted as zero if \( m_n < m \).

Using now the Minkowski inequality, the stationarity of the \( X_i \)'s and (2.21), one gets

\[
E[|\sum_{i=1}^{n} X_i|^m] \leq \sum_{i=1}^{\infty} E[|X_i|^m] \quad \text{and} \quad \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} E[|X_i|^m | X_j|^m] \leq \sum_{i=1}^{\infty} |X_i|^m + \sum_{j=1}^{\infty} |X_j|^m.
\]

\[
\leq K_1 \left[ \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} \nu_{r+1-j} 2^{(r+1-j)R(m)(\log n)^{K_2} + 2^{(r+1-j)R(m)}} \right]^m \leq K_1 \left[ \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} \nu_{r+1-j} 2^{(r+1-j)R(m)(\log n)^{K_2} + 2^{(r+1-j)R(m)}} \right]^m.
\]

(2.23)

3. Proof of the theorem. We need only prove (1.2) as (1.3) is then immediate.

Recall we have already assumed \( \mu = 0 \) and \( \sigma = 1 \). Let \( p = p_n = [n^q] \), \( q = q_n = [n^{(1/2 + \beta - \alpha)}] \); \( \alpha \) and \( \beta (0 < \beta < \alpha < 1) \) will be chosen later; \([x]\) denotes the largest integer \( \leq x \). Then, \( n = k(p + q) + r \), \( 0 \leq r < p + q \).

The sample sum \( S_n \) is then partitioned into

\[
S_n' = U_n + R_n + T_n,
\]

where

\[
U_n = \sum_{i=1}^{k} \xi_i; \quad \xi_i = \xi_{n,i} = \sum_{j=i}^{i+j} \nu_{r+1-j} 2^{(r+1-j)R(m)(\log n)^{K_2} + 2^{(r+1-j)R(m)}} \leq k;
\]

\[
R_n = \sum_{i=1}^{k} \eta_i; \quad \eta_i = \eta_{n,i} = \sum_{j=i}^{i+j} \nu_{r+1-j} 2^{(r+1-j)R(m)(\log n)^{K_2} + 2^{(r+1-j)R(m)}} \leq k;
\]

\[
T_n = T_{n,r} = \sum_{i=k+1}^{(p+q)+1} X_j \quad \text{if} \quad r \geq 1; \quad T_n = 0 \quad \text{if} \quad r = 0.
\]

We now prove three lemmas which imply (2.3), and thus prove the theorem. Define \( \zeta_n = (\log n)^{q-1}, n \geq 2 \).

**Lemma 3.** For any \( c > 0 \), under (1.4)—(1.6),

\[
P \left[ U_n > (c \pm 2\zeta_n(n \log n)^{K_2}) \right] \sim (2\pi c^2 \log n)^{-1/2} n^{-c^2}.
\]

**Lemma 4.** For any \( c > 0 \), under (1.4)—(1.6),

\[
P \left[ R_n > \zeta_n(n \log n)^{K_2} \right] = o(n^{-c^2}(\log n)^{-2}) = P \left[ R_n < -\zeta_n(n \log n)^{K_2} \right].
\]

**Lemma 5.** For any \( c > 0 \), under (1.4)—(1.6),

\[
P \left[ T_n > \zeta_n(n \log n)^{K_2} \right] = o(n^{-c^2}(\log n)^{-2}) = P \left[ T_n < -\zeta_n(n \log n)^{K_2} \right].
\]

Further, for \( \zeta_n < \frac{1}{2} c \), \( n \) large, one has

\[
P \left[ U_n > (c \pm 2\zeta_n(n \log n)^{K_2}) \right] - P \left[ \hat{R}_n < -\zeta_n(n \log n)^{K_2} \right]
\]

\[
- P \left[ T_n < -\zeta_n(n \log n)^{K_2} \right]
\]

\[
\leq P \left[ S_n' > c(n \log n)^{K_2} \right]
\]

\[
\leq P \left[ U_n > (c - 2\zeta_n(n \log n)^{K_2}) \right] + P \left[ R_n > \zeta_n(n \log n)^{K_2} \right]
\]

\[
+ P \left[ T_n > \zeta_n(n \log n)^{K_2} \right].
\]
Dividing both sides of (3.8) by \((2\pi^3 \log n)^{-1} n^{-1+\varepsilon}\), and then letting \(n \to \infty\), one gets (2.3).

**Proof of Lemma 3.** For later reference, note that

\[
\log [n^{-i} e^{2 \zeta n} n^{k+2}] = \frac{1}{3} [e^2 - (e^2 + 2 \zeta n)] \log n = O_\nu (\log n)^{-\varepsilon}, \quad \nu > 0,
\]

where \(O_\nu\) denotes exact order. By \(M_n = O_\nu(a_n)\), we mean that there exist constants \(K_1(> 0)\) and \(K_2(> 0)\), and a positive integer \(n_0\) such that \(K_1 \leq |M_n|/a_n \leq K_2\) for \(n \geq n_0\). We have used (3.9) in deriving (3.10).

Defining \(\xi_i' = p^{-1} \xi_i(1 \leq i \leq k)\) and \(c_n = (c + 2 \zeta n)(n/(k p))^{1/2}(\log n/\log k)^{1/2}\),

\[
P(U_n > (c + 2 \zeta n)(n \log n)^{1/2}) = P(\sum \xi_i' > c_n(k \log k)^{1/2}).
\]

Define now

\[
\xi_{i0} = \begin{cases} 
\xi_i' & \text{if } |\xi_i'| \leq (k/\log k)^{1/2}, \\
0 & \text{otherwise};
\end{cases} \quad 1 \leq i \leq k.
\]

Using Lemma 2, one has

\[
|P(\sum \xi_i' > c_n(k \log k)^{1/2}) - P(\sum \xi_{i0} > c_n(k \log k)^{1/2})| \
\leq kP(|\xi_i'| > (k/\log k)^{1/2}) \leq k(k/\log k)^{-1+\varepsilon} E[|\xi_i'|^m] \
\leq k(k/\log k)^{-1+\varepsilon} \varepsilon_{K_1}(p n^{m-\varepsilon - 2-\delta}/(\log n)^{\varepsilon_2} + p^{m/2}) \
\leq K_2 n (n/kp)^{-m/2} n^{-1+\varepsilon - 1/2}(\log n)^{\varepsilon_2} + k(-1+\varepsilon)(\log n)^{\varepsilon_2} \
\leq K_1 n^{-\varepsilon/2}(\log n)^{-1/2},
\]

for some \(\eta > 0\), and for large \(n\), if \((1 - \alpha)(m - 2) > c^2\), i.e.,

\[
(13) \quad \alpha < 1 - c^2/(m - 2).
\]

Thus, to prove the lemma, it suffices to show that

\[
P(\sum \xi_{i0} > c_n(k \log k)^{1/2}) \sim (2\pi^3 \log n)^{-1} n^{-1+\varepsilon}.
\]

To prove (3.14), one proceeds in analogy with Rubin and Sethuraman (1965). Let \(F_k(x) = P(\sum \xi_{i0} \leq x)\), \(g_k(\theta) = \exp[\theta x] dF_k(x)\), and \(dG_k(\theta) = \exp(\theta x) dF_k(x)\). Then it is possible to express the LHS of (3.14) as

\[
g_k(c_n(k \log k)^{1/2}) \int_{-\infty}^{\infty} \exp(-c_n(k \log k)^{1/2} x) dG_k(c_n(k \log k)^{1/2})(x).
\]

Define now

\[
f_n = E[\exp(c_n(k \log k)^{1/2} \xi_{i0})];
\]

\[
m_n = f_n^{-1} E[\xi_{i0} \exp(c_n(k \log k)^{1/2} \xi_{i0})];
\]

\[
nm_n^2 + \sigma_n^2 = f_n^{-1} E[\xi_{i0}^2 \exp(c_n(k \log k)^{1/2} \xi_{i0})].
\]

Then after some simple manipulations, it is possible to write

\[
P(\sum \xi_{i0} > c_n(k \log k)^{1/2}) = A_k \int_{\sigma_k} \exp(-c_n \xi^2) d\Pi_{k+c_n}(\xi),
\]
where \( A_k = A_k(c_n) = E[\exp\{c_n(\log k/k)^i \sum \xi_{10}\}] \exp(-c_n(k \log k)^i m_n) \), \( B_k = B_k(c_n) = k^{-i}c_n^{-1}(c_n(k \log k)^i - km_n) \), \( C_k = c_n m_n(k \log k)^i \), \( \Pi_k(c_n) = C_k \exp(\sum \log(k/k_i)) \). Next we obtain some useful estimates of \( A_k, B_k \) and \( C_k \). Towards this end, first note that using Lemma 2

\[
E[\xi_{10}]^3 \leq E[\xi_{10}]^3 \leq K_i p^{-i}(p^{1}\ +\ pn^{-i(\alpha-1)}) \leq K_i^{1-i}k^{-\gamma},
\]

\(1 \leq i \leq k\), for some \(\gamma > 0\). Also,

\[
E[\xi_{10}]^2 \leq E[\xi_{10}^2] \leq K_i.
\]

Further, using (3.11) and (3.19),

\[
|E(\xi_{10})| \leq (k/\log k)^{-i}E[\xi_{10}]^3 \leq K_i k^{-i-\gamma},
\]

for some \(\gamma > 0\). Next we show that

\[
|E[\xi_{10}^2 - 1]| \leq K_i k^{-\gamma}
\]

for some \(\gamma > 0\).

This is accomplished in several steps. First note that using (3.19)

\[
|E[\xi_{10}^2 - 1]| = E\left[|\xi_{10} I(\xi_{10} > (k/\log k)^{1/2})|\right] \leq (k/\log k)^{-i}E[\xi_{10}]^3 \leq K_i k^{-i-\gamma}
\]

for some \(\gamma > 0\).

Also, using the stationarity of the \(X_i\)'s, Lemma 1, page 170 of Billingsley (1968), (1.4), and \(\sum_{j=1}^{\gamma} \phi^i(j) \leq K_i\) (which follows from (1.5)), one gets

\[
|p^{-1}E(\sum_{i=1}^{\gamma} X_i)^{2} - \xi_{10}^2| \leq p^{-1}pE(X_i^2 - X_i^2) + 2 \sum_{j=1}^{\gamma} (p - j)E(X_i X_{i+j} - X_i X_{i+j})
\]

\[
\leq E[X_i^2 I_{[\xi_{10} > (k/\log k)^{1/2}]}) + 4 \sum_{j=1}^{\gamma} |E(X_i I_{[\xi_{10} > (k/\log k)^{1/2}]})| + \sum_{j=1}^{\gamma} \phi^i(j)E(X_i I_{[\Xi_{10} > (k/\log k)^{1/2}]})E(X_i I_{[\Xi_{10} > (k/\log k)^{1/2}]})
\]

\[
\leq n^{-i(\alpha-\delta)}K_i n^{-i(\alpha-\delta)} + 8 \sum_{j=1}^{\gamma} \phi^i(j)E(X_i^2 I_{[\Xi_{10} > (k/\log k)^{1/2}]}) + 4 \sum_{j=1}^{\gamma} \phi^i(j)E(X_i I_{[\Xi_{10} > (k/\log k)^{1/2}]}) + 6pE[X_i I_{[\Xi_{10} > (k/\log k)^{1/2}]})^2
\]

\[
\leq K_i n^{-i(\alpha-\delta)} + K_i n^{-i(\alpha-\delta)} + K_i n^{-i(\alpha-\delta)} + K_i n^{-i(\alpha-\delta)} + K_i n^{-i(\alpha-\delta)} + K_i n^{-i(\alpha-\delta)} \leq K_i n^{-\gamma}
\]

for some \(\gamma > 0\).

In view of (3.23) and (3.24), \(EX_i = 0\) and \(\sigma^2 = 1\), (3.22) will now follow from the following lemma.

**Lemma 6.** Under (1.4)—(1.6), \(|p^{-1}V(\sum_{i=1}^{\gamma} X_i) - \sigma^2| \leq K_i n^{-\gamma}\) for some \(\gamma > 0\).

**Proof.**

\[
|\sigma^2 - p^{-1}V(\sum_{i=1}^{\gamma} X_i)| \leq 2(\sum_{j=1}^{\gamma} j|p^{-1}\text{Cov}(X_i, X_{i+j})| + \sum_{j=1}^{\gamma} |\text{Cov}(X_i, X_{i+j})|)
\]

\[
\leq 2(\sum_{j=1}^{\gamma} j|p^{-1}\phi^i(\alpha+\beta+\gamma)(j)(E[X_i])^{(\alpha+\beta+\gamma)/(\alpha+\beta+\gamma)}\times (E[X_i])^{(\alpha+\beta+\gamma)/(\alpha+\beta+\gamma)} + \sum_{j=1}^{\gamma} \phi^i(\alpha+\beta+\gamma)(j)
\]

\[
\times (E[X_i])^{(\alpha+\beta+\gamma)/(\alpha+\beta+\gamma)}\times (E[X_i])^{(\alpha+\beta+\gamma)/(\alpha+\beta+\gamma)}(E[X_i])^{(\alpha+\beta+\gamma)/(\alpha+\beta+\gamma)}
\]

\[
\leq K_i p^{-1} \sum_{j=1}^{\gamma} j^{-\alpha-\beta-\gamma}(\alpha+\beta+\gamma) + \sum_{j=1}^{\gamma} j^{-1}(\alpha+\beta+\gamma)
\]

\[
\leq K_i p^{-\gamma} \leq K_i n^{-\gamma},
\]
for some $\gamma > 0$, since (1.5) and the monotonicity of the $\phi(j)$'s give $\phi(j) \leq K_j j^{-\gamma}$.

With the usual expansion of $\exp(x)$ around $x = 0$, it follows from (3.20)—(3.22) that

$$\left| f_n - 1 - \frac{1}{2} c_n^2 \log(k/k) \right| \leq K_j k^{-1-\gamma} \quad \text{for some } \gamma > 0. \quad (3.25)$$

Further, from (3.16)—(3.17) and (3.20)—(3.22),

$$m_n = c_n (\log(k/k)) \xi \xi \in (\xi) + O(k^{1-\gamma})$$
$$= c_n (\log(k/k)) + O(k^{1-\gamma}) \quad \text{for some } \gamma > 0; \quad (3.26)$$

$$m_n^2 + a_n^2 = 1 + O(k^{-\gamma}), \quad \text{for some } \gamma > 0. \quad (3.27)$$

Hence, $a_n^2 = 1 + O(k^{-\gamma})$ for some $\gamma > 0$, and one has

$$B_n = k^{-1} (1 + O(k^{-\gamma})) = O(k^{-\gamma}). \quad (3.28)$$

Also $n/kp = 1 + (kq + r)/(kp) = 1 + O(n^{-\gamma})$, for some $\gamma > 0$. Hence,

$$C_n = c_n a_n (\log k)^{1/2} = c(\log n)^{(1/2)(1 + O(\log n)^{-\gamma})}. \quad (3.29)$$

We prove next

$$A_n = n^{1/2} (1 + O(\log n)^{-\gamma}) \quad \text{for some } \gamma > 0. \quad (3.30)$$

Since $-c_n (\log k)^{1/2} = -c_n^2 \log k + O(k^{-1} \log k) = -c^2 \log n + O((\log n)^{-\gamma})$ for some $\gamma > 0$, it suffices to show that

$$E[\exp(c_n (\log k/k)^{1/2} \xi_0)] = n^{i\theta} [1 + O(\log n)^{-\gamma}], \quad (3.31)$$

for some $\gamma > 0$. Note that in view of (3.25), $k^{-1} f_n = \frac{1}{2} c_n^2 \log k + O(k^{-\gamma}) = \frac{1}{2} c^2 \log n + O((\log n)^{-\gamma})$. Hence,

$$f_n^k = n^{i\theta} (1 + O(\log n)^{-\gamma}), \quad \text{for some } \gamma > 0. \quad (3.32)$$

In view of (3.31), in proving (3.30), it suffices to show that

$$|E[\exp(c_n (\log k/k)^{1/2} \xi_0)] - f_n^k| = O(n^{i\theta - \gamma}), \quad \text{for some } \gamma > 0. \quad (3.33)$$

To prove (3.32), we make repeated use of the inequality (20.28) in Billingsley (1968), page 171. The inequality is due to Ibragimov (1962).

Using stationarity and $|\xi_0| \leq (k/k)^{1/2}$, one gets

$$|E[\exp(c_n (\log k/k)^{1/2} \xi_0)] - f_n^k| \leq |E[\exp(c_n (\log k/k)^{1/2} \xi_0)] - E[\exp(c_n (\log k/k)^{1/2} \xi_0)] f_n^k|$$
$$+ f_n^k |E[\exp(c_n (\log k/k)^{1/2} \xi_0)] - f_n^k|$$
$$- E[\exp(c_n (\log k/k)^{1/2} \xi_0)] f_n^k + \cdots$$
$$+ f_n^{k-1} |E[\exp(c_n (\log k/k)^{1/2} \xi_0)] - f_n^k|$$

$$\leq 2 \phi(q) \exp(2c_n)$$
$$\{ \sum_{j=1}^{k-1} f_n^{k-1-j} E[\exp(c_n (\log k/k)^{1/2} \xi_0)] + f_n^{k-1} \}.$$

Again, repeated application of (20.28) of Billingsley gives

$$E[\exp(c_n (\log k/k)^{1/2} \xi_0)]$$
$$\leq |f_n + 2 \phi(q) \exp(c_n) E[\exp(c_n (\log k/k)^{1/2} \xi_0)]|$$
$$\leq |f_n + 2 \phi(q) \exp(c_n)|^j, \quad j = 2, \ldots, k - 1. \quad (3.34)$$
It follows now from (3.33) that
\begin{align}
|E(\exp(c_n (\log k/k)^{\alpha} \sum \xi_{i_0}) - f_n^h)| \\
\leq 2k\phi(q) \exp(2c_n)[f_n + 2\phi(q) \exp(c_n)]^{b-1}
\leq K_1 q^{-2k/\alpha} \exp(k\log k/k) + K_1(k^{-1-\gamma} + q^{-2})^{b-1},
\end{align}
where (3.25) and (1.5). Note that for large \( n \) the RHS of (3.35) is majorized by
\( K_1 n^{-2b+1-\alpha} \exp[k \log (1 + 2c_n^2 \log k/k)] \) for some \( \gamma > 0 \) provided
\( 2\beta > 1 - \alpha \). The above is again majorized by
\( K_1 n^{-2b+1-\alpha} \exp(2c_n^2 \log k + K_1 k^{-1-\gamma}) = O(n^{b-2b+1-\alpha}) \). Choose now \( \alpha = 1 - (c_1 + \delta)/(m - 2) \), \( \beta = 1 - (c_1 + 2\delta)/(m - 2) \), where \( m > 3c_1 + 5\delta + 2 \). Then, (3.13) is satisfied, \( 0 < \beta < \alpha < 1 \) and \( 2\beta > 1 - \alpha \). Also, \( 1 - \alpha - 2\beta = -2 + (3c_1 + 5\delta)/(m - 2) < -1 \). This proves (3.32). Finally we show
\begin{equation}
\sup |\Pi_{h,n}(z) - \Phi(z)| = O((\log n)^{-1}).
\end{equation}

Assume for the moment that this is true. To see how the proof is concluded first note that in view of (3.18) and \( A_{h} = n^{-1}O(1 + O(\log n)^{-1}) \), for some \( \gamma > 0 \), it suffices to show that
\begin{equation}
|\Pi_{h}^{\alpha} \exp(-C_{h} z) d\Pi_{h,n}(z) - \Pi_{h}^{\alpha} \exp(-C_{h} z) d\Phi(z)| = O((\log n)^{-1}).
\end{equation}

From (3.28), (2.29) and (3.36) one has
\begin{equation}
|\Pi_{h}^{\alpha} \exp(-C_{h} z) d\Pi_{h,n}(z) - \Pi_{h}^{\alpha} \exp(-C_{h} z) d\Phi(z)| = O((\log n)^{-1}).
\end{equation}

Hence, one need only prove that
\begin{equation}
\Pi_{h}^{\alpha} \exp(-C_{h} z) d\Phi(z) \sim (2\pi c_1^2 \log n)^{-1}.
\end{equation}

It is easy to see that \(|\Pi_{h}^{\alpha} \exp(-C_{h} z) d\Phi(z) - \Pi_{h}^{\alpha} \exp(-C_{h} z) d\Phi(z)| \leq K_{1} |B_{h}| = O(n^{-\gamma}) \) for some \( \gamma > 0 \) (using (3.28)). Also,
\begin{equation}
\Pi_{h}^{\alpha} \exp(-C_{h} z) d\Phi(z) = \Pi_{h}^{\alpha} (2\pi)^{-1} \exp(-\frac{1}{2}(z + C_{h})^{2} + \frac{1}{2}C_{h} z) dz
= (2\pi)^{-1} [1 - \Phi(C_{h})] / N(C_{h}) \sim (2\pi)^{-1} C_{h}^{-1} \sim (2\pi c_1^2 \log n)^{-1}.
\end{equation}

It remains to prove (3.36). If \( H(x) = P(\xi_{i_0} \leq x) \) and \( dH_{n}(x) = \exp(c_n (\log k/k)^{\alpha} x) dH(x) / \sum \exp(c_n (\log k/k)^{\alpha} x) dH(x) \), it follows that \( m_{n} \) and \( \sigma_{n} \) are the mean and the variance of \( \xi_{i_0} \) wrt the conjugate distribution \( H_{n} \). If \( \xi_{i_0}^{n} (i = 1, \ldots, k) \) are independent rv's each having the same conjugate distribution \( H_{n} \) as \( \xi_{i_0} \), Lemma 2, page 171 of Billingsley (1968) gives
\begin{equation}
|E_{H_{n}}(\Pi_{n}^{b} \exp[it(\xi_{i_0} - m_{n})k^{-1}\sigma_{n}^{-1}]) - [\Pi_{n}^{b} E_{H_{n}}[\exp[it(\xi_{i_0} - m_{n})k^{-1}\sigma_{n}^{-1}] - 2k\phi(q) \leq K_1 k^{-2} = O(n^{-\gamma}) \) for some \( \gamma > 0 \); \end{equation}
in the above \( E_{H_{n}} \) denotes expectation wrt \( H_{n} \). Let now \( F_{1}(x) \) and \( F_{2}(x) \) denote the respective df's of \( \sum (\xi_{i_0} - m_{n})k^{-1}\sigma_{n}^{-1} \) and \( \sum (\xi_{i_0} - m_{n})k^{-1}\sigma_{n}^{-1} \) when the \( \xi_{i_0} \)'s and \( \xi_{i_0}^{n} \)'s have df \( H_{n}(x) \). Let \( f_{1}, f_{2}, \) and \( f_{3} \) denote the characteristic functions corresponding to \( \Phi, F_{1} \) and \( F_{2} \). Define \( \beta_{n} = E_{H_{n}}[\xi_{i_0}^{n}]^{2} \). It follows from (3.21) and
(3.25) that $\beta_s = O(k^{1-r})$ for some $\gamma > 0$. Define $T_n = \sigma_a^{\alpha}h_a^{-1}k^b$ so that $T_n^{-1} = O(k^{1-r})$. Proceeding in analogy with Theorem 5.1 of Feller (1966, pages 515–516), one gets for $|t| \leq T_n$,

$$
|f_s(t) - f_0(t)| \leq K\beta_{s}\sigma_a^{-3-k}a^{-k}t^{1+R^2} = O(n^{-\gamma})t^{1+R^2}.
$$

Using now Esseen’s lemma (see, e.g., Feller (1966), page 512), one gets for all real $z$,

$$
|F_s(z) - \Phi(z)| \leq K\left[\int_{[\tau_a]}^{\tau_a}|f_s(t) - f_{0}(t)||t|^{-1}dt + T_n^{-1}\right]
$$

Again $|f_s(t) - f_{0}(t)| < K(\log n)^{-1}t$ for $|t| < (\log n)^{-1}$ (since the corresponding variables admit moments of order higher than two). Using this and (3.40), one gets for large $n$,

$$
\int_{[\tau_a]}^{\tau_a}|f_s(t) - f_{0}(t)||t|^{-1}dt = \int_{[t<\log n)}^{\tau_a}|f_s(t) - f_{0}(t)||t|^{-1}dt
$$

noting that $\log T_n = O(\log n)$; (3.41), (3.42) and the above now lead to

$$
\sup_{z}|F_s(z) - \Phi(z)| = O((\log n)^{-1}).
$$

But $F_s$ is the same as $\Pi_n\sigma_n$. This proves (3.36). The proof of Lemma 3 is now complete.

PROOF OF LEMMA 4. We need only prove that $P(R_n > \zeta_n(n \log n)^{1}) = o(n^{-1+\alpha}(\log n)^{-1})$ as the proof of the other quality in (3.6) is analogous. To this end, first define,

$$
\eta_{i0} = \eta_i, \quad \text{if} \quad |\eta_i| \leq n^{1/2}\log n^{1/2},
$$

$$
= 0, \quad \text{otherwise}.
$$

Then, using the stationarity of the $\eta_i$’s, and Lemma 2,

$$
|P(R_n > \zeta_n(n \log n)^{1}) - P(\sum^\infty_1 \eta_{i0} > \zeta_n(n \log n)^{1})|
$$

$$
\leq kP(\eta_i > n^{1/2}\log n^{1/2}) \leq kE|\eta_i|^{m-n^{2/3}
$$

$$
\leq K_n[k\eta_i^{m-n^{2/3}}\log n]^{K_n} + q^{1/2}n^{m-n^{2/3}}\log n^{m-n^{2/3}}
$$

But

$$
1 - \alpha - \frac{m}{2}(1 - \beta) = \frac{c^2 + \delta}{m - 2} - \frac{m}{2} \frac{c^2 + 2\delta}{m - 2}
$$

$$
= \frac{-(m - 2)c^2 - 2(m - 1)\delta}{2(m - 2)} = -\frac{1}{2}c^2 - \frac{m - 1}{m - 2}\delta.
$$

Hence, it follows from (3.44) that

$$
|P(R_n > \zeta_n(n \log n)^{1/2}) - P(\sum^\infty_1 \eta_{i0} > \zeta_n(n \log n)^{1/2})|
$$

$$
\leq K_n n^{-1+\alpha - \gamma} \quad \text{for some} \quad \gamma > 0.
$$
Thus, it suffices to show that
\[ P(\sum_{i}^{d} \eta_{i0} > \zeta_\alpha(n \log n)^{\frac{1}{2}}) = o(n^{-d}(\log n)^{-\frac{1}{2}}). \]

Note first that
\[
\begin{aligned}
P(\sum_{i}^{d} \eta_{i0} > \zeta_\alpha(n \log n)^{\frac{1}{2}}) &\leq \inf_{\theta > 0} \left[ \exp(-\theta n^{-1}(\log n)^{\frac{3}{2} + \epsilon} \zeta_\alpha(n \log n)^{\frac{1}{2}}) \times E(\exp(\theta n^{-1}(\log n)^{\frac{3}{2} + \epsilon} \sum_{i}^{d} \eta_{i0})) \right].
\end{aligned}
\]

Using arguments similar to (3.33)--(3.35), it follows that
\[
|E(\exp(\theta n^{-1}(\log n)^{\frac{3}{2} + \epsilon} \sum_{i}^{d} \eta_{i0})) - h_{n}^k| \leq 2k\phi(p) \exp(2\theta)[h_{n} + 2\theta \phi(p)]^{k-1},
\]
where \( h_{n} = E(\exp(\theta n^{-1}(\log n)^{\frac{3}{2} + \epsilon} \eta_{i0})) \). By a Taylor expansion we have
\[
h_{n} \leq 1 + \theta n^{-1}(\log n)^{\frac{3}{2} + \epsilon} |E \eta_{i0}| + \frac{1}{2} \theta^2 n^{-1}(\log n)^{\frac{5}{2} + \epsilon} E(\eta_{i0}) \exp(\theta).
\]

But, using Lemma 2,
\[
|E(\eta_{i0})| = |E(\mathbb{I}_{[\eta_{i0} > n^{\frac{1}{2}}(\log n)^{\frac{3}{2} + \epsilon}]})| \leq n^{-1}(\log n)^{\frac{3}{2} + \epsilon} E(\eta_{i0}) \leq K_1 q n^{-1}(\log n)^{\frac{3}{2} + \epsilon}.
\]

Also,
\[
|E \eta_{i0}| = q |EX_i| \leq q E[|X_i| \mathbb{I}_{[|X_i| > n^{\frac{1}{2}}]})] \leq q(n^{-1})^{\frac{3}{2} + \epsilon + \frac{1}{2}} E|X_i|^{\frac{3}{2} + \epsilon + \frac{1}{2}} \leq K_1 q n^{-1}(\log n)^{\frac{3}{2} + \epsilon}.
\]

Thus,
\[
|E \eta_{i0}| \leq K_1 q n^{-1}(\log n)^{\frac{3}{2} + \epsilon}.
\]

Also, using Lemma 2, \( E \eta_{i0}^2 \leq E \eta_{i0}^2 \leq K_1 q \). Thus one has
\[
(k - 1) \log (h_{n} + 2\phi(p) \exp(\theta)) \leq (k - 1) \log (1 + q n^{-1}(\log n)^{\frac{3}{2} + \epsilon} + K_1 p^{-2})
\]
\[
\leq k[q n^{-1}(\log n)^{\frac{3}{2} + \epsilon} + K_1 n^{-2\alpha}] \leq K_1[n^{3\alpha}(\log n)^{3\alpha + 2\epsilon} + n^{1-3\alpha}]
\]
\[
= K_1[n^{3\alpha}(\log n)^{3\alpha + 2\epsilon} + n^{-2(\alpha + \epsilon - \alpha)(\alpha + \epsilon)}] \leq K_1 n^{-\gamma} \text{ for some } \gamma > 0,
\]

since \( m > 3\alpha^2 + 5\delta + 2 \). One has now from (3.47), for large \( n \),
\[
|E(\exp(\theta n^{-1}(\log n)^{\frac{3}{2} + \epsilon} \sum_{i}^{d} \eta_{i0})) - h_{n}^k| \leq K_1 n^{-3\alpha} \exp(K_1 n^{-\gamma}) \leq K_1 n^{-\gamma}
\]
for some \( \gamma > 0 \). Also, (3.50) leads to \( k \log h_{n} \leq K_1 n^{-\gamma} \) for some \( \gamma > 0 \). So, \( h_{n}^k \leq 1 + K_1 n^{-\gamma} \) for some \( \gamma > 0 \). Thus, from (3.51) and the above,
\[
\text{the RHS of (3.46) } \leq \inf_{\theta > 0} n^{-\theta}(1 + K_1 n^{-\gamma}) < K_1 n^{-3\alpha - \gamma} \text{ for some } \gamma > 0;
\]
(3.46), and hence the lemma follows.

**Proof of Lemma 5.** Using \( r \leq p + q \), Markov’s inequality and Lemma 2,
\[
P(|T_n| > \zeta_\alpha(n \log n)^{\frac{1}{2}})
\]
\[
\leq \{n^k(\log n)^{-\frac{3}{2} - \frac{1}{2}}E|T_n|^m \}
\]
\[
\leq n^{-m/2}(\log n)^{m(1 + \epsilon)} K_1[(p + q)n^{\frac{1}{2} - 2\alpha - \epsilon}(\log n)^{\frac{1}{2}} + (p + q)^{m/2}].
\]
Since \((p + q)n = O(n^{-1+\gamma}) = O(n^{-\alpha^2/(m-1)})\), it follows from (3.53) that \(P(|T_n| > \zeta_n(n \log n)^{\gamma}) \leq K_n n^{-\alpha^2 \gamma} \) for some \(\gamma > 0\) by a suitable choice of \(m (> 0)\). This completes the proof of the theorem.

Acknowledgments. We thank the referee for his constructive criticisms and for pointing out an error in an earlier version of the paper. Thanks are also due to Professors D. L. Hanson for several helpful comments, Dean Isaacson for some useful discussions and Richard A. Groeneveld for his careful reading of the manuscript and for suggesting several stylistic improvements.

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