

L_1 BOUNDS FOR ASYMPTOTIC NORMALITY OF m -DEPENDENT SUMS USING STEIN'S TECHNIQUE¹

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In a recent paper, C. Stein has given a new, direct technique for bounding the error of the normal approximation to the distribution of a sum of dependent random variables, assuming the variables form a stationary sequence with eighth moments. In the present paper we give two L_1 bounds on this error for an arbitrary m -dependent sequence with second moments.

1. Introduction, notation and results. Let X_1, X_2, \dots be random variables with $EX_k = 0$, $EX_k^2 < \infty$. Set

$$\begin{aligned} s_n^2 &= E(\sum_1^n X_k)^2, & F_n(x) &= P(\sum_1^n X_k \leq s_n x) \\ L_{\alpha n} &= \sum_1^n E|X_k|^\alpha / s_n^\alpha & \text{and} & \quad R_n = R_{3n} + R_{2n}, \\ R_{3n} &= \sum_1^n E(|X_k|^3 / s_n^3; |X_k| \leq s_n), \\ R_{2n} &= \sum_1^n E(X_k^2 / s_n^2; |X_k| > s_n). \end{aligned}$$

Let \mathcal{N} denote the standard normal distribution with density

$$\varphi(x) = (2\pi)^{-1/2} \exp(-x^2/2).$$

If X_1, X_2, \dots are independent, Feller (1968b) shows that $\|F_n - \mathcal{N}\|_\infty \leq 6R_n$; and the author (1973) shows $\|F_n - \mathcal{N}\|_1 < 72R_n$.

In this paper we extend the L_1 result to include m -dependent random variables. To be precise, we prove

THEOREM A. *Suppose X_1, X_2, \dots are m -dependent. Set $M_3 = (2m + 1)(4m + 1)(6m + 1)$. Then*

$$\|F_n - \mathcal{N}\|_1 \leq 13M_3(L_{3n} + L_{4n}).$$

(See Theorem 3.9 for a somewhat sharper result.)

Since this is of little value when fourth moments fail to exist, we truncate and rescale to obtain

THEOREM B. *Suppose X_1, X_2, \dots are m -dependent. Set $M_3 = (2m + 1)(4m + 1)(6m + 1)$. Then*

$$\|F_n - \mathcal{N}\|_1 \leq 336M_3R_n.$$

(See Theorem 4.6 and the paragraph following.)

It should be noted that if X_1, X_2, \dots are independent, identically distributed

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then $m = 0$ and we have the “classical rate”

$$\|F_n - \mathcal{N}\|_1 \leq \min \{13(A_3/n^3 + A_4/n), 336A_{2+\alpha}/n^{\alpha/2}\}$$

where $A_\beta = E|X_1|^\beta / (EX_1^2)^{\beta/2}$.

The procedure used to prove Theorem 1 is the extremely interesting new technique of Stein (1972), which is direct and makes no use of characteristic functions. It is hoped that our use of this technique to prove an L_1 result will stimulate study of this method. It is noteworthy that the L_1 result proceeds much more simply than the L_∞ result of Stein, and that, in the non independent, identically distributed case, we as yet have found no way to prove the “correct” L_∞ result. All that we have is the trivial

THEOREM C. *Let X_1, X_2, \dots be m -dependent. Set $M_3 = (2m + 1)(4m + 1)(6m + 1)$. Then*

$$\|F_n - \mathcal{N}\|_\infty \leq 17[M_3 R_n]^{1/2}.$$

In the original version of this paper we included results for the case when X_1, X_2, \dots satisfy certain mixing conditions. We omit these, however, since in their present rough form they obscure the beautiful technique of Stein, and since they are too crude to yield Theorem A when specialized to the m -dependent case.

We wish to thank Professor Stein for stimulating discussion during the summer of 1972, especially for helping with the bound of E_{11} below which improves our original version and leads to the “correct rate” in Theorem A.

In the final section of this paper we give an example which shows that the so-called “correct rate” for m -dependent variables may be far from correct.

2. Stein’s equation. Stein (1969) gives an interesting characterization of the standard normal random variable: W has distribution \mathcal{N} iff $E[f'(W) - Wf(W)] = 0$ for sufficiently many functions f . Thus, to see if W is approximately normal, check the magnitude of $E[f'(W) - Wf(W)]$. It is exactly this that Stein does.

To proceed let us first verify the above characterization. This is done most simply by defining the *normal transform* of any function h in $\mathbf{B} = \{h : R \rightarrow R \mid h \text{ is bounded, measurable}\}$:

$$\begin{aligned} h^{\mathcal{N}}(x) &= \int_{-\infty}^x [h(y) - \nu h]_{\mathcal{N}}(y) dy_{\mathcal{N}}(x) \\ &= - \int_x^\infty [h(y) - \nu h]_{\mathcal{N}}(y) dy_{\mathcal{N}}(x), \end{aligned}$$

where $\nu h = \int_{-\infty}^\infty h(y)_{\mathcal{N}}(y) dy = Eh(\mathcal{N})$. (We shall consistently use \mathcal{N} for both the standard normal distribution and its corresponding random variable on any appropriate probability space.) Further, define

$$h^{\mathcal{N}^d}(x) = xh^{\mathcal{N}}(x) + h(x) - \nu h.$$

(2.1) **PROPOSITION.** *For any h in \mathbf{B} and any finite a, b , we have*

$$\|h^{\mathcal{N}}\|_\infty \leq 4\|h\|_\infty, \quad \|h^{\mathcal{N}^d}\|_\infty \leq 4\|h\|_\infty$$

and $\int_a^b h^{\mathcal{N}^d}(x) dx = h^{\mathcal{N}}(b) - h^{\mathcal{N}}(a)$.

PROOF. When $x < 0$, $|h^{\nu}(x)| \leq 2\|h\|_{\infty} \mathcal{N}(x)/\nu(x)$, and $\mathcal{N}(x)/\nu(x) \leq \min(1/|x|, 1/2\nu(x)) \leq 2$ (for the last use $x = -\frac{1}{2}$ and for the first notice that $\int_{-\infty}^x -y\nu(y) dy / (-x) = \nu(x)/(-x)$). This same reasoning shows that $|x|\mathcal{N}(x)/\nu(x) \leq 1$, and hence $\|h^{\nu^d}\|_{\infty} \leq 4\|h\|_{\infty}$. The last assertion follows by using Fubini's theorem. \square

The characterization of normal is now obvious: for each h in \mathbf{B} set $f = h^{\nu}$; if $W = \mathcal{N}$ then $E[f'(\mathcal{N}) - \mathcal{N}f(\mathcal{N})] = \nu h - \nu h = 0$, and conversely if for all h in \mathbf{B} , $0 = E(f'(W) - Wf(W)) = Eh(W) - Eh(\mathcal{N})$ then $W = \mathcal{N}$.

The next step in Stein's technique is to bound $E[f'(W) - Wf(W)] = Eh(W) - Eh(\mathcal{N})$, h in \mathbf{B} , $f = h^{\nu}$, $f' = h^{\nu^d}$. This is done in great generality in Stein (1972), but here we shall deal only with the m -dependent case.

Fix $n > m$, and from now on *Assume*

$$\begin{aligned} \{X_i\}_1^n & \text{ is } m\text{-dependent, i.e. for all } i < n - m \\ \sigma\{X_1, \dots, X_i\} & \text{ is independent of } \sigma\{X_{i+m+1}, \dots, X_n\} \end{aligned}$$

and use the Notation

$$\begin{aligned} A_i &= X_i/s_n, & B_i &= \sum_{|j-i| \leq m} A_j, & C_i &= \sum_{|j-i| \leq 2m} A_j \\ D_i &= \sum_{|j-i| \leq 3m} A_j, & Q_i &= A_i B_i - EA_i B_i, & S &= \sum A_j \\ b_i &= S - B_i, & c_i &= S - C_i, & d_i &= S - D_i, \\ f &= h^{\nu}, & f' &= h^{\nu^d} & \text{ for } h & \text{ in } \mathbf{B}. \end{aligned}$$

Summation is always taken over the indicated range and between 1 and n .

Under the assumptions it is clear that the entries in the pair's (A_i, b_i) , (B_i, c_i) , (C_i, d_i) and (Q_i, c_i) are independent.

(2.2) PROPOSITION (Stein's equation). *Suppose $L_{4n} < \infty$. Under the above assumptions for each h in \mathbf{B} , $Eh(S) - Eh(\mathcal{N}) = \sum_1^4 E_k$, where $E_k = E_k(h)$ is given by*

$$\begin{aligned} E_1 &= \sum E[f'(S) - f'(c_i)]Q_i \\ E_2 &= \frac{1}{2} \sum EA_i B_i^2 f(S) \\ E_3 &= \sum EA_i \int_s^{b_i} z \int_s^z f'(u) du dz \\ E_4 &= \sum EA_i \int_s^{b_i} [h(z) - h(S)] dz. \end{aligned}$$

PROOF. Observe that

$$Eh(S) - Eh(\mathcal{N}) = E[f'(S) - Sf(S)] = E\{f'(S) + \sum A_i [f(b_i) - f(S)]\}.$$

Now write $f(b) - f(s) = \int_s^b [f'(z) - f'(s)] dz + (b - s)f'(s)$ and $f'(z) - f'(s) = (z - s)f'(s) + z \int_s^z f''(u) du + h(z) - h(s)$.

Formal substitution produces terms E_2 , E_3 and E_4 and would force $E_1 = Ef'(S)[1 - \sum A_i B_i]$. The expression for E_1 now follows because m -dependence implies both that $1 = ES^2 = \sum EA_i B_i$ and that c_i and Q_i are independent. Under the assumption that L_{4n} is finite it is easy to show that each E_k is finite (more details are given in the succeeding propositions), and the formal substitution is justified. \square

3. Proof of Theorem A. We would have a nice L_∞ Berry–Esséen result for m -dependent random variables if we could show $|E_k(h)| \leq \text{const.} \cdot \|h\|_\infty(L_{3n} + L_{4n})$. We cannot do this; nor do we see how this can be done replacing $(L_{3n} + L_{4n})$ by something of the form $\sum_{k=3}^8 L_{kn}$, as in Stein (1972).

But we can do something for an L_1 bound, $\|F_n - \mathcal{N}\|_1$. In order to obtain such a bound take $h = h_x = I_{(-\infty, x]}$ so that $Eh(S) - Eh(\mathcal{N}) = F_n(x) - \mathcal{N}(x) = \sum_1^4 E_k(h_x)$. We show how to bound

$$E_{k1} = \int_{-\infty}^{\infty} |E_k(h_x)| dx .$$

First we must prove something like (2.1).

(3.1) **PROPOSITION.** Set $h = h_x = I_{(-\infty, x]}$, $f_x = h^{\mathcal{N}}$, $f'_x = h^{\mathcal{N}'}$. For each real y we have

$$\int_{-\infty}^{\infty} |f_x(y)| dx = 1 , \quad \int_{-\infty}^{\infty} |f'_x(y)| dx \leq 3 .$$

PROOF. From the definition of the normal transform it is clear that

$$\begin{aligned} f_x(y) &= \mathcal{N}(y)[1 - \mathcal{N}(x)]/\mathcal{N}(x) && \text{if } y \leq x \\ &= \mathcal{N}(x)[1 - \mathcal{N}(y)]/\mathcal{N}(x) && \text{if } y \geq x . \end{aligned}$$

The first equality now follows by direct calculation, using Fubini’s theorem and expressing $\mathcal{N}(y) = \int_{-\infty}^y \mathcal{N}(u) du$. Let $L_1(y)$ denote the left side of the second inequality. Similarly it is shown that

$$L_1(y) = 2\mathcal{N}(y)[1 + y\mathcal{N}(y)]/[1 - y(1 - \mathcal{N}(y))] .$$

If use is made of the fact (Feller (1968a), page 193) $p_{2k+1}(x) \leq [1 - \mathcal{N}(x)]/\mathcal{N}(x) \leq p_{2k}(x)$, $k = 0, 1, \dots$, $x > 0$, where $p_k(x) = x^{-1} + \sum_1^k (-1)^j [1 \cdot 3 \cdot \dots \cdot (2j - 1)]x^{-(2j+1)}$, it is easy to show that $L_1(y) \leq 3$, $|yL_1(y)| \leq 3$ and $\lim_{|y| \rightarrow \infty} |yL_1(y)| = 2$. \square

Before proceeding to give bounds for E_{k1} we state an extension of an inequality of Loève (1963), page 155:

(3.2) **PROPOSITION** (c_r -inequality). Given random variables U_1, \dots, U_k , set $c_{rk} = \max(1, kr^{-1})$. Then $E|U_1 + \dots + U_k|^r \leq c_{rk} \sum_1^k E|U_j|^r$.

To get the best bounds we randomize, as in Stein (1972). Let J be a random variable independent of all X_1, \dots, X_n and let J be uniform on the set of indices $\{1, \dots, n\}$. Now consider the randomized random variables A_J, B_J, Q_J .

(3.3) **PROPOSITION.** For $\alpha \geq 1$ we have $E|A_J|^\alpha \leq L_{\alpha n}/n$, $E|B_J|^\alpha \leq (2m + 1)^\alpha L_{\alpha n}/n$, $E|Q_J|^\alpha \leq 2^\alpha E|A_J B_J|^\alpha$.

PROOF. Note first that $E|A_J|^\alpha = n^{-1} \sum E|A_j|^\alpha = n^{-1} L_{\alpha n}$. Next, by (3.2), $E|B_J|^\alpha = \sum E|B_j|^\alpha/n \leq (2m + 1)^{\alpha-1} \sum_j \sum_{|k-j| \leq 2m} E|A_k|^\alpha/n \leq (2m + 1)^\alpha L_{\alpha n}/n$. Finally, again by (3.2) and also the moment inequality, $E|Q_J|^\alpha \leq 2^{\alpha-1} [E|A_J B_J|^\alpha + E|q_J|^\alpha]$, where $|q_j|^\alpha = |E(A_j B_j)|^\alpha \leq E|A_j B_j|^\alpha$. \square

Similar results are true for $E|C_J|^\alpha$, $E|C_J - D_J|^\alpha, \dots$ and these will be used without further comment. Another inequality typical of what will be used

below is

(3.4) PROPOSITION. Set $K_J = C_J - D_J$. Then $nE|C_J K_J Q_J| \leq 2(2m + 1)(4m + 1)(2m)L_{4n}$.

PROOF. Use a three term Hölder inequality to get

$$\begin{aligned} \|C_J K_J Q_J\|_1 &\leq \|C_J\|_4 \|K_J\|_4 \|Q_J\|_2 \leq 2\|C_J\|_4 \|K_J\|_4 \|A_J\|_4 \|B_J\|_4 \\ &\leq 2[(4m + 1)^4 (L_{4n}/n)(2m)^4 (L_{4n}/n)(L_{4n}/n)(2m + 1)^4 L_{4n}/n]^{\frac{1}{2}}. \quad \square \end{aligned}$$

We are now in a position to find appropriate bounds for E_{k1} .

(3.5) PROPOSITION.

$$E_{11} \leq 10(2m + 1)(4m + 1)L_{3n} + 6(2m + 1)(4m + 1)(8m + 1)L_{4n}.$$

PROOF. We may certainly assume $L_{4n} < \infty$, else there is nothing to prove. From the definition of f_x' we have $E_1(h_x) = nE[f_x'(S) - f_x'(c_J)]Q_J = nE\{C_J f_x(S)Q_J + c_J Q_J \int_{c_J}^S f_x'(u) du + [h_x(S) - h_x(c_J)]Q_J\}$. Now $\int_{-\infty}^{\infty} |h_x(S) - h_x(c_J)| dx = |C_J|$, and (3.1) implies

$$E_{11} \leq nE|C_J Q_J| + 3nE|c_J C_J Q_J| + nE|C_J Q_J|.$$

As in (3.4), $E|C_J Q_J| \leq \|C_J\|_3 \|Q_J\|_{\frac{3}{2}} \leq 2\|C_J\|_3 \|A_J\|_3 \|B_J\|_3 \leq 2(2m + 1)(4m + 1)L_{3n}/n$. Next $c_J = d_J + (D_J - C_J)$ and d_J is independent of $C_J Q_J$. Thus

$$E|c_J C_J Q_J| \leq E|d_J C_J Q_J| + E|K_J C_J Q_J|, \quad K_J = D_J - C_J.$$

Finally, since $E|d_J| \leq E|S| + E|D_J| \leq 1 + \delta_j$, $\delta_j = E|D_J|$, we have $E|d_J C_J Q_J| = n^{-1} \sum E|C_J Q_J| E|d_J| \leq E|C_J Q_J| + E|\delta_j C_J Q_J|$. But $E|\delta_j|^4 \leq n^{-1} \sum E|D_J|^4 \leq (6m + 1)^4 L_{4n}/n$, and we argue as in (3.4) to obtain $E|\delta_j C_J Q_J| \leq \|\delta_j\|_4 \|C_J\|_4 \|Q_J\|_2 \leq 2(6m + 1)(4m + 1)(2m + 1)L_{4n}/n$. Adding all of the estimates gives the result. \square

(3.6) PROPOSITION. $E_{21} \leq \frac{1}{2}(2m + 1)^2 L_{3n}$.

PROOF. $2E_{21} \leq nE|A_J B_J^2| \leq n\|A_J\|_3 \|B_J^2\|_{\frac{3}{2}} = n\|A_J\|_3 \|B_J\|_3^2 \leq n(L_{3n}/n)^{\frac{1}{2}}((2m + 1)^3 L_{3n}/n)^{\frac{3}{2}}$. \square

(3.7) PROPOSITION. $E_{31} \leq \frac{3}{2}(2m + 1)^2 L_{3n} + 3[\frac{5}{6}(2m + 1)^3 + \frac{1}{2}(2m + 1)^2(6m + 1)]L_{4n}$.

PROOF. Since $|z| \leq |z - S| + |S|$ we have

$$\begin{aligned} \frac{1}{3}E_{31} &\leq nE|A_J \int_S^{b_J} |z| |z - S| dz| \\ &\leq \frac{n}{3} E|A_J B_J^3| + \frac{n}{2} E|A_J B_J^2 S|. \end{aligned}$$

Write $S = B_J + c_J + K_J$, $K_J = C_J - B_J$ to get $\frac{1}{3}E_{31} \leq \frac{5}{6}H_1 + \frac{1}{2}H_2 + \frac{1}{2}H_3$, where $H_1 = nE|A_J B_J^3| \leq n\|A_J\|_4 \|B_J\|_4^3 \leq (2m + 1)^3 L_{4n}$, $H_2 = nE|A_J B_J^2 K_J| \leq (2m + 1)^2(2m)L_{4n}$ and (as in the proof of (3.5)) $H_3 = nE|A_J B_J^2 c_J| \leq nE|A_J B_J^2| + nE|\gamma_J A_J B_J^2| \leq (2m + 1)^2 L_{3n} + (2m + 1)^2(4m + 1)L_{4n}$, where $\gamma_j = E|C_j|$. \square

(3.8) PROPOSITION. $E_{41} \leq \frac{1}{2}(2m + 1)^2 L_{3n}$.

PROOF. $\int_{-\infty}^{\infty} |h_x(z) - h_x(S)| dx = |z - S|$, and thus $E_{41} \leq (n/2)E|A_J B_J^2|$. \square

Addition gives the next result, of which Theorem A is a simple corollary.

(3.9) THEOREM. *Suppose X_1, X_2, \dots are m -dependent. Then for all n*

$$\begin{aligned} \|F_n - \mathcal{N}\|_1 &\leq (12.5)(2m + 1)(3.6m + 1)L_{3n} \\ &\quad + 10(2m + 1)(22m^2 + 9.4m + 1)L_{4n}. \end{aligned}$$

Let us point out the problem in getting L_∞ bounds by the above methods. For the L_1 bound we encountered $\int |h_x(z) - h_x(S)| dx = |z - S|$, $h_x = I_{(-\infty, x]}$ both in E_{11} and E_{41} ; in the L_∞ case this is instead $\sup_x |h_x(z) - h_x(S)| = 1$. This is exactly the problem. Some bounds can be given, but they lead to the square root of the ‘‘correct bound’’ and a proof of Theorem C which is not as nice as our geometric proof using Theorem B. (See the last paragraphs of Section 4.)

4. Proof of Theorem B and C. We prove Theorem B by giving an L_1 bound for the error committed by truncation. We also give results which apply to the L_∞ case. To do this we first prove

(4.1) PROPOSITION. *Let U and V be random variables with distributions G and H . Then*

$$\|G - H\|_1 \leq E|U - V|, \quad \|G - H\|_\infty \leq P(U \neq V).$$

PROOF. The L_∞ inequality follows from the identity $G(x) - H(x) = P(U \leq x, U \neq V) - P(V \leq x, U \neq V)$. For the L_1 inequality define $I_a = I_{(-\infty, a]}$, so that $\|G - H\|_1 = \int_{-\infty}^{\infty} |E I_a(U) - E I_a(V)| da \leq E \int_{-\infty}^{\infty} |I_a(U) - I_a(V)| da = E|U - V|$. (This trick now has been used three times.) \square

(4.2) PROPOSITION. *Let $a > 0$ and define $\mathcal{N}'_a(x) = \mathcal{N}(ax)$. Then $\|\mathcal{N}'_a - \mathcal{N}\|_1 \leq \frac{4}{5}|1 - a^{-1}|$ and $\|\mathcal{N}'_a - \mathcal{N}\|_\infty \leq \frac{4}{5}|1 - a^{-1}|$.*

PROOF. Let U have distribution \mathcal{N} and $V = U/a$ have distribution \mathcal{N}'_a . Then $E|U - V| = |1 - a^{-1}|E|U| \leq \frac{4}{5}|1 - a^{-1}|$ and the L_1 part follows from (4.1). For the L_∞ part, first take $a \geq 1$ and notice that $|\mathcal{N}(ax) - \mathcal{N}(x)| = |\int_{x/a}^{ax} \varphi(y) dy| \leq |a - 1| |\varphi(x)|$. Since $|\varphi(x)|$ is maximal at $|x| = 1$ and $\varphi(1) < \frac{4}{5}$ we have the result, for $\|\mathcal{N}'_a - \mathcal{N}\|_\infty = \|\mathcal{N} - \mathcal{N}'_{a^{-1}}\|_\infty$. \square

Let us use these in the following truncation setup. Set $\tau^2 = s_n^2 = E(\sum X_k)^2$; define $X_k^\tau = X_k I(|X_k| \leq \tau)$, and

$$Y_k = X_k^\tau - EX_k^\tau, \quad Z_k = X_k - Y_k.$$

Set $T = \sum Y_k$, $t_n^2 = ET^2$, $r = s_n/t_n$. Finally, write $F_n - \mathcal{N} = H_1 + H_2 + H_3$, $H_1(x) = F_n(x) - P(T \leq s_n x)$, $H_2(x) = P(T \leq t_n rx) - \mathcal{N}(rx)$ and $H_3(x) = \mathcal{N}'_r(x) - \mathcal{N}(x)$.

We will bound H_k using $L_{\alpha n}(Z) = \sum E|Z_k|^\alpha/s_n^\alpha$ and $L_{\alpha n}(Y) = \sum E|Y_k|^\alpha/s_n^\alpha$.

From (4.1) follows

(4.3) PROPOSITION. $\|H_1\|_1 \leq L_{1n}(Z)$.

Define $M_k = \prod_{j=1}^k (2jm + 1)$.

From (3.9) and the change of variable $y = rx$ follows

(4.4) PROPOSITION. *We have*

$$\|H_2\|_1 \leq (12.5)r^2 M_2 L_{3n}(Y) + 10r^3 M_3 L_{4n}(Y).$$

Clearly $\|H_3\|_1 \leq \frac{4}{5}|1 - r^{-2}|$, because of (4.2). Let us bound this.

(4.5) PROPOSITION. $|1 - r^{-2}| \leq M_1[4L_{1n}(Z) + L_{2n}(Z)]$.

PROOF. Notice that $1 - r^{-2} = (s_n^2 - t_n^2)/s_n^2$ and that $s_n^2 - t_n^2 = \text{Var}(\sum X_k) - \text{Var}(T) = \text{Var}(\sum Z_k) + 2 \text{Cov}(T, \sum Z_k)$. Because of m -dependence $\text{Cov}(Z_k, Z_j) = 0 = \text{Cov}(Y_k, Z_j)$ if $|j - k| > m$. In general, $|2 \text{Cov}(Z_k, Z_j)| \leq EZ_k^2 + EZ_j^2$, and $|\text{Cov}(Y_k, Z_j)| \leq E|Y_k Z_j| \leq 2s_n E|Z_j|$. The result follows by writing $s_n^2 - t_n^2 = \sum [EZ_k^2 + 2 \text{Cov}(Y_k, Z_k)] + \sum_{k=1}^m \sum_{j=1}^{n-k} [2 \text{Cov}(Z_j, Z_{k+j}) + 2 \text{Cov}(Z_j, Y_{k+j}) + 2 \text{Cov}(Z_{k+j}, Y_j)]$. \square

Combining these we prove

(4.6) THEOREM. $\|F_n - \mathcal{N}\|_1 \leq 14(K_1 + K_2)$, where $K_1 = M_1[4L_{1n}(Z) + L_{2n}(Z)]$ and $K_2 = M_2 L_{3n}(Y) + M_3 L_{4n}(Y)$.

PROOF. By (4.1) we know that $\|F_n - \mathcal{N}\|_1 \leq E|S - \mathcal{N}| \leq \|S - \mathcal{N}\|_2 = [\text{Var}(S) + \text{Var} \mathcal{N}]^{1/2} \leq 2^{1/2}$. Thus we may assume $K_1 \leq 2^{1/2}/14$, and from $1 - r^{-2} \leq K_1$ it follows that $r^2 \leq 1.12$, $r^3 \leq 1.2$. Substitution of these values in the bound for $\|H_2\|_1$ and addition of the bounds for $\|H_k\|_1$ give the result. \square

To prove Theorem B use the c_r -inequality (3.2) and the moment inequality to see that $L_{3n}(Y) \leq 8R_{3n}$, $L_{4n}(Y) \leq 16R_{3n}$, while $L_{1n}(Z) \leq 2R_{2n}$, $L_{2n}(Z) \leq R_{2n}$.

To prove Theorem C, if $\|F_n - \mathcal{N}\|_\infty = h$, draw a triangle with height h and base $5h/2$ completely between the graphs of F_n and \mathcal{N} . This can be done since \mathcal{N} has maximum slope $\approx(0) < \frac{2}{5}$. This triangle has area equal $5h^2/4 \leq \|F_n - \mathcal{N}\|_1$ and we see that $\|F_n - \mathcal{N}\|_\infty \leq 2[\|F_n - \mathcal{N}\|_1/5]^{1/2} \leq 16.5(M_3 R_n)^{1/2}$.

5. An example. Suppose X_1, X_2, \dots are identically distributed and 1-dependent and that $EX_1^4 < \infty$. Then (3.9) implies that

$$\|F_n - \mathcal{N}\|_1 \leq 1000Kn(s_n^{-3} + s_n^{-4}) = B_n$$

where $E|X_1|^3, EX_1^4 \leq K < \infty$. We will construct a sequence so that

$$B_n > Cn^{1/2}, \quad \|F_n - \mathcal{N}\|_1 \leq Cn^{-1/2}$$

for some constant $C, 0 < C < \infty$. This will show that while our bounds in terms of L_{an} look effective, in fact they are not the proper generalization of the classical Berry-Esséen theorem.

In order to construct such a sequence we merely have to choose the X_i in such a way that

$$s_n^2 = E(\sum_1^n X_k)^2 \sim n^{1/2}$$

while $S_n = \sum_1^n X_k$ consists only of essentially $n^{1/2}$ independent, identically distributed (i.i.d.) terms.

To do this let Y_1, Y_2, \dots be i.i.d. with $EY_1 = 0$, $EY_1^2 = 1$, $EY_1^4 = K < \infty$. Notice that if $X_{2k-1} = Y_k$, $X_{2k} = -Y_k$ then $s_n^2 = 0$ if n even, $= 1$ if n odd. If we do not force quite so much cancellation we have the result we want.

More precisely, define X_k inductively as follows. Set $X_1 = Y_1$, $X_2 = Y_2$. After k terms are defined, when $X_k = Y_j$, define $X_{k+1} = -Y_j$ if $s_k^2 > k^{\frac{1}{2}}$ and $X_{k+1} = +Y_{j+1}$ if $s_k^2 \leq k^{\frac{1}{2}}$; on the other hand, if $X_k = -Y_j$, just define $X_{k+1} = Y_{j+1}$. A graph makes it clear that $|s_n^2 - n^{\frac{1}{2}}| \leq 2$. Therefore $\sum_1^n X_k$ contains approximately $n^{\frac{1}{2}}$ nonzero i.i.d. terms. If we apply (3.9) with $m = 0$ to this i.i.d. sequence we get $\|F_n - \mathcal{N}\|_1 \leq C'n^{\frac{1}{2}}(n^{-\frac{1}{2}} + n^{-1}) < Cn^{-\frac{1}{2}}$. The result now follows from the definition of B_n .

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