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CRAMÉR TYPE LARGE DEVIATIONS FOR GENERALIZED RANK STATISTICS[1]

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A Cramér type large deviation theorem is proved under alternatives as well as under hypothesis for the generalized linear rank statistic which includes as special cases (unsigned) linear rank statistics, signed linear rank statistics, linear combination of functions of order statistics, and a rank combinatorial statistic.

1. Introduction. Let $X_{N1}, X_{N2}, \cdots, X_{NN}$ be independent r.v.'s (random variables) and let g be a real valued measurable function such that $X_{Nj}^g \equiv g(X_{Nj})$, $1 \le j \le N$, has a c.d.f. (cumulative distribution function) G_{Nj}. We introduce the generalized linear rank statistic

$$T_N = \textstyle\sum_{j=1}^N c_{Nj} a_{NR_{Nj}^g}(X_{Nj}) \tag{1.1}$$

where $\{c_{Nj}: 1 \le j \le N\}$ is an array of regression constants; $\{a_{Nj}(\cdot): 1 \le j \le N\}$ is an array of known real functions (called scores); and $R_{Nj}^g = \sum_{k=1}^N u(X_{Nj}^g - X_{Nk}^g)$ is a generalized rank of X_{Nj}^g among $\{X_{Nk}^g: 1 \le k \le N\}$ where $u(x) = 1$ if $x \ge 0$ and $u(x) = 0$ otherwise. We assume that the scores $a_{Nj}(\cdot)$, $1 \le j \le N$ are generated by a nonconstant score generating function $\phi(s, t)$, $0 < s < 1$, $-\infty < t < \infty$, in either of the following two ways;

$$a_{Nj}(t) = \phi(EU_{N:j}, t), \quad j = 1, 2, \cdots, N \text{ (approximate scores)} \tag{1.2}$$

$$a_{Nj}(t) = E\phi(U_{N:j}, t) \quad j = 1, 2, \cdots, N \text{ (exact scores)} \tag{1.3}$$

where $U_{N:j}$ is the jth order statistic in a random sample of size N from the uniform distribution over (0, 1). To avoid the trivialities, we assume that $\sum_{j=1}^N |c_{Nj}| > 0$.

When $g(x) = x$ and $\phi(x, y) = \tilde{\phi}(x)$, the statistic (1.1) reduces to the (unsigned) linear rank statistic

$$T_N = \textstyle\sum_{j=1}^N c_{Nj} \tilde{a}_{NR_{Nj}} \tag{1.4}$$

where R_{Nj}, $1 \le j \le N$, is the rank of X_{Nj} among $\{X_{Nk}: 1 \le k \le N\}$ and $\tilde{a}_{Nj}$'s are usual scores of constants. On the other hand, when $g(x) = |x|$ and $\phi(x, y) = \tilde{\phi}(x)\operatorname{sgn} y$, where $\operatorname{sgn} y = 2u(y) - 1$, the statistic (1.1) reduces to the signed linear

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rank statistic

$$T_N = \sum_{j=1}^{N} c_{Nj} \tilde{a}_{NR^+_{Nj}} \operatorname{sgn} X_{Nj} \tag{1.5}$$

where R^+_{Nj}, $1 \le j \le N$, is the rank of $|X_{Nj}|$ among $\{|X_{Nk}|: 1 \le k \le N\}$.

Also note that when $c_{N1} = \cdots = c_{NN} = N^{-1}$, $g(x) = x$, $\phi(x, y) = \tilde{\phi}(x)\psi(y)$, and the underlying c.d.f.'s G_{Nj} are continuous, the statistic (1.1) reduces to the linear combination of functions of order statistics

$$T_N = (1/N) \sum_{j=1}^{N} \tilde{a}_{NR_{Nj}} \psi(X_{Nj}) = (1/N) \sum_{j=1}^{N} \tilde{a}_{Nj} \psi(X_{N:j}) \tag{1.6}$$

where $X_{N:j}$ is the jth order statistic among $\{X_{Nk}, 1 \le k \le N\}$. Furthermore, denoting $Y_{jk} = c_{Nj} a_{Nk}(X_{Nj})$, $1 \le j, k \le N$, we can rewrite (1.1) as

$$T_N = \sum_{j=1}^{N} Y_{jR^*_{Nj}} \tag{1.7}$$

which is a rank combinatorial statistic. For these four different types of statistics, several authors have investigated problems concerning asymptotic normality, rate of convergence to normality and higher order expansions. For a review, the reader is referred to Hájek (1962, 1968), Hušková (1970, 1977, 1979), Jurečková and Puri (1975), Bergström and Puri (1977), Puri and Seoh (1984a, b, c), Does (1982), Puri and Wu (1983) (for statistics of the type (1.4) and (1.5)); Shorack (1969, 1972), Stigler (1974), Bjerve (1977) and Helmers (1977, 1980, 1981) (for statistics of the type (1.6)); Hoeffding (1951), Motoo (1956), von Bahr (1976) and Ho and Chen (1978) (for the statistics of the type (1.7)) and the papers cited therein, among others.

In recent years, there has been a great upsurge of activity in the theory of large deviations initiated by Cramér (1938) and studied in detail by Petrov (1975) for the case of independent summands. Large deviation probabilities for U-statistics were obtained by Malevich and Abdalimov (1979) while the corresponding results for the case of the statistic (1.4) were studied by Kallenberg (1982). But these results are restrictive in the sense that the observations $X_{N1}, X_{N2}, \cdots, X_{NN}$ are identically distributed and the underlying distribution function is *continuous* (Kallenberg, 1982). In addition, Robinson (1977) has dealt with the case of large deviations for samples from finite populations.

In this paper we shall be concerned with the relative error of the normal approximation to the distribution of the (properly normalized) generalized linear rank statistic (1.1) under general alternatives, i.e., assuming only that the observations are independent (*not necessarily identically distributed*) *and without assuming the continuity of the underlying distribution functions*. The results obtained not only include the results of Kallenberg (1982) as a special case, but also provide the large deviation theory for the statistics of the type (1.5), (1.6), and (1.7) which to the best of our knowledge has not been considered in the generality of the present paper.

2. Assumptions and main theorem. Throughout this paper we make the following assumptions.

ASSUMPTION (A). The variance of T_N satisfy

$$\liminf \tau_N^2 = \liminf \operatorname{Var} T_N > 0.$$

ASSUMPTION (B). The regression constants $c_{N1}, c_{N2}, \cdots, c_{NN}$ satisfy

$$\sum_{j=1}^N c_{Nj}^2 = 1, \quad \max_{1\le j\le N} |c_{Nj}| \le A_1 N^{-1/3}, \qquad \sum_{j=1}^N |c_{Nj}|^3 \le A_2 N^{-1/2},$$

where A_1 and A_2 are absolute constants.

ASSUMPTION (C). The score generating function $\phi(s, t)$ is differentiable with respect to its first argument s such that its first partial derivatives $\phi_1(s, t) = \partial\phi(s, t)/\partial s$ satisfy Lipschitz's condition of order one with respect to s, i.e., there is a constant Δ such that for any $x, y \in (0, 1)$

$$\sup_{-\infty<t<\infty} |\phi_1(x, t) - \phi_1(y, t)| \le \Delta |x - y|. \tag{2.1}$$

(Note that the normal scores statistic does not satisfy this assumption).

We now introduce some notations. Let S° denote a r.v. S centered at its expectation, i.e. $S^\circ = S - ES$. Denote

$$\rho_{Nj} = R_{Nj}^g/(N + 1), \qquad \rho_{Njj} = E(\rho_{Nj} \mid X_j). \tag{2.2}$$

Furthermore, we shall use the r.v. $\hat{S}_N$ (Hájek's projection), as an approximation of the statistic T_N, defined by

$$\hat{S}_N = \sum_{\ell=1}^N E(S_N \mid X_\ell) - (N - 1)ES_N \tag{2.3}$$

where S_N is the first two terms of Taylor's expansion of T_N with approximate scores, i.e.,

$$S_N = \sum_{j=1}^N c_{Nj}\{\phi(\rho_{Njj}, X_{Nj}) + (\rho_{Nj} - \rho_{Njj})\phi_1(\rho_{Njj}, X_{Nj})\}. \tag{2.4}$$

Let $\Phi(\cdot)$ denote the standard normal c.d.f. and put $\hat{\sigma}_N^2 = \operatorname{var} \hat{S}_N$. Then

THEOREM 2.1. *Under assumptions* (A), (B) *and* (C), *uniformly in the region* $0 < x \le \rho_N N^{1/6}$, $\rho_N = o(1)$, *we have as* $N \to \infty$

$$P(T_N - ET_N > \tau_N x)[1 - \Phi(x)]^{-1} = 1 + o(1) \tag{2.5}$$

which remains true if we replace τ_N by $\hat{\sigma}_N$.

REMARK 2.1. Note that the result of Kallenberg (1982) which deals with the statistic (1.4) and holds for the case of iid r.v.'s is a special case of our result, but we impose somewhat stronger assumption on the score generating function. This is due to our generalized statistic (1.1) and weaker assumptions on underlying distributions. However, for the case of iid r.v.'s, the results of the above theorem hold under relatively weaker assumptions (see Remark 4.1).

From now on, we shall suppress the subscript N in c_{Nj}, a_{Nj}, R^g_{Nj}, X_{Nj}, ρ_{Nj}, ρ_{Njj}, etc., whenever it causes no confusion.

3. Preliminaries. In this section, we derive bounds on the $2r$th moments (r is any positive integer) for statistics defined in Lemma 3.1–3.5. These bounds play an important role in this paper and, since the method of their derivation depends heavily on that of Bickel (1974), we will give only the brief outline of the proofs.

Let $\{Y_j\}_{j=1}^{\infty}$ be a sequence of r.v.'s and $\{d_j\}_{j=1}^{\infty}$ a sequence of real numbers. Then we have the following lemmas.

LEMMA 3.1. *Let Z_j, $j \geq 1$, be r.v.'s of the form $Z_j = g_j(Y_1, Y_2, \cdots, Y_j)$ such that for $j \geq 2$, $E(Z_j \mid Y_1, Y_2, \cdots, Y_{j-1}) = 0$. If the sequence $\{d_j\}$ is nonincreasing in absolute values, then for any positive integers r and ℓ,*

$$E(\textstyle\sum_{j=1}^{\ell} d_j Z_j)^{2r} \leq (4e)^r \, (\sum_{j=1}^{\ell} d_j^2)^r \, r^r \max_{i\leq j\leq \ell} EZ_j^{2r}.$$

LEMMA 3.2. *Let Y_j, $j \geq 1$, be independent r.v.'s and let $\tilde{V}_{jk}$ be r.v.'s of the form $\tilde{V}_{jk} = g_{jk}(Y_j, Y_k)$, $1 \leq j, k < \infty$, such that for any j and k, $j \neq k$, $E(\tilde{V}_{jk} \mid Y_j) = E(\tilde{V}_{jk} \mid Y_k) = 0$. Then, for any positive integers ℓ and r,*

$$E[(1/\ell) \textstyle\sum_{j=1}^{\ell} \sum_{k\neq j}^{\ell} d_j \tilde{V}_{jk}]^{2r} \leq (4e)^{2r} (\sum_{j=1}^{\ell} d_j^2)^r (2r)^{2r} \ell^{-r} \max_{1\leq j,k\leq \ell; j\neq k} E\tilde{V}_{jk}^{2r}.$$

REMARK 3.1. Lemmas 3.1 and 3.2 are generalizations of Lemmas 1 and 4, respectively, of Hušková (1979). (We may point out that, in Hušková (1979), the proof of Lemma 4 is incorrect in its application of Lemma 1, especially in deriving (23) and (24) of her paper). Lemma 3.2 is also a generalization, as well as an improvement, of Lemma 2.2 of Bergström and Puri (1977).

PROOF OF LEMMA 3.1. For $r \geq \ell$, the proof follows by applying Hölder's inequality to $(\sum_{j=1}^{\ell} d_j Z_j)^{2r}$ and some routine computations. For $r \leq \ell$, the proof follows by induction on ℓ with r fixed.

PROOF OF LEMMA 3.2. Since the assumptions and the conclusion of this lemma are invariant under simultaneous permutation of d_j's and Y_j's, we may, without loss of generality, assume that $|d_1| \geq |d_2| \geq \cdots \geq |d_\ell|$.

Define $Z_1 = \tilde{Z}_1 = 0$, $Z_j = \sum_{k=1}^{j-1} \tilde{V}_{jk}$ and $\tilde{Z}_j = \sum_{k=1}^{j-1} d_k \tilde{V}_{jk}$, $2 \leq j \leq \ell$, so that $(1/\ell) \sum_{j=1}^{\ell} \sum_{k\neq j}^{\ell} d_j \tilde{V}_{jk} = (1/\ell)(\sum_{j=1}^{\ell} d_j Z_j + \sum_{j=1}^{\ell} \tilde{Z}_j) = \hat{V}_\ell$, say. The proof then follows by using the following facts:

(a) $E\hat{V}_\ell^{2r} \leq \ell^{-2r} 2^{2r-1} (4e)^r r^r \{(\sum_{j=1}^{\ell} d_j^2)^r \max_{1\leq j\leq \ell} EZ_j^{2r} + \ell^r \max_{1\leq j\leq \ell} E\tilde{Z}_j^{2r}\}$.

(b) $EZ_j^{2r} \leq (4e)^r (j-1)^r r^r \max_{1\leq k\leq \ell; k\neq j} EV_{jk}^{2r}$ and

(c) $E\tilde{Z}_j^{2r} \leq (4e)^r (\sum_{k=1}^{j-1} d_k^2)^r r^r \max_{1\leq k\leq \ell; k\neq j} EV_{kj}^{2r}$.

LEMMA 3.3. *Let T_N and S_N be defined by* (1.1) *and* (2.4). *Then for any positive* *integer* $r \geq 1$, $E(T_N^0 - S_N^0)^{2r} \leq (16e\Delta)^{2r}(2r)^{2r}N^{-r}$.

PROOF. Denote $H_{jk} = u(X_j^g - X_k^g) - G_k(X_j^g)$, $1 \leq j, k \leq N$. Then, we have

$$\rho_j - \rho_{jj} = (N+1)^{-1} \sum_{k\neq j}^N H_{jk} \tag{3.1}$$

and

$$S_N = \sum_{j=1}^N c_j\{\phi(\rho_{jj}, X_j) + (N+1)^{-1} \sum_{k\neq j}^N H_{jk}\phi_1(\rho_{jj}, X_j)\}. \tag{3.2}$$

First consider the statistic T_N with approximate scores. Then, by Taylor's expansion, for some $0 \leq \lambda \leq 1$,

$$T_N = \sum_{j=1}^N c_j\{\phi(\rho_{jj}, X_j) + (\rho_j - \rho_{jj})\phi_1(\lambda\rho_{jj} + (1-\lambda)\rho_j, X_j)\}. \tag{3.3}$$

Using (3.2) and (3.3) along with assumptions (B) and (C), we obtain

$$E(T_N - S_N)^{2r} \leq \Delta^{2r}N^{r-1} \sum_{j=1}^N E(\rho_j - \rho_{jj})^{4r}. \tag{3.4}$$

For each j, conditionally given X_j^g, $\rho_j - \rho_{jj}$ is the sum of independent r.v.'s with zero means and thus we may apply Lemma 3.1 to obtain that for any integer $r \geq 1$,

$$E(\rho_j - \rho_{jj})^{4r} \leq (4e)^{2r}(2r)^{2r}N^{-2r}. \tag{3.5}$$

The proof follows using (3.4) and (3.5).

We now consider the exact scores. Let $\tilde{T}_N$ be the statistic (1.1) with exact scores and put for $1 \leq j \leq N$, $\tilde{a}_j(t) = E\phi(U_{N:j}, t)$ to distinguish it from the statistic T_N with approximate scores. The proof then follows by using $|E\phi(U_{N:j}, t) - \phi(j/(N+1), t)| \leq \Delta N^{-1}$, $E(\tilde{T}_N^0 - T_N^0)^{2r} \leq (2\Delta)^{2r}N^{-r}$, $E(T_N^0 - S_N^0)^{2r} \leq 2^{2r}E(T_N - S_N)^{2r} \leq (8e\Delta)^{2r}(2r)^{2r}N^{-r}$ and routine computations.

LEMMA 3.4. *Let $\hat{S}_N$ and S_N be defined by* (2.3) *and* (2.4). *Then for any integer $r \geq 1$, $E(S_N^0 - \hat{S}_N^0)^{2r} \leq (8e\|\phi_1\|)^{2r}(2r)^{2r}N^{-r}$ where*

$$\|\phi_1\| = \sup_{0<s<1}\sup_{-\infty<t<\infty}|\phi_1(s, t)|.$$

PROOF. The proof follows by using Lemma 3.2 and noting that

$$\begin{aligned} E(S_N \mid X_\ell) = {} & c_\ell\phi(\rho_{\ell\ell}, X_\ell) + ES_N - c_\ell E\phi(\rho_{\ell\ell}, X_\ell) \\ & + (N+1)^{-1} \sum_{j=1}^N c_j E\{H_{j\ell}\phi_1(\rho_{jj}, X_j) \mid X_\ell\}, \end{aligned} \tag{3.6}$$

and

$$S_N^0 - \hat{S}_N^0 = (N+1)^{-1} \sum_{j=1}^N \sum_{k\neq j}^N c_j[H_{jk}\phi_1(\rho_{jj}, X_j) - E\{H_{jk}\phi_1(\rho_{jj}, X_j) \mid X_k\}].$$

LEMMA 3.5. *Let T_N and $\hat{S}_N$ be defined by* (1.1) *and* (2.3). *Then, for any real $r \geq 1/2$,*

$$E(T_N^0 - \hat{S}_N^0)^{2r} \leq (64e(\Delta + \|\phi_1\|))^{2r}(2r)^{2r}N^{-r}. \tag{3.7}$$

PROOF. Denote by $[x]$, the smallest integer $\geq x$. Then, applying Hölder's inequality,

$$[E(T_N^0 - \hat{S}_N^0)^{2r}]^{[r]/r} \le E(T_N^0 - \hat{S}_N^0)^{2[r]}.$$

Since, by the C_r-inequality,

$$\text{the right-hand side term} \le 2^{2[r]-1}[E(T_N^0 - S_N^0)^{2[r]} + E(S_N^0 - \hat{S}_N^0)^{2[r]}],$$

we have

$$[E(T_N^0 - \hat{S}_N^0)^{2r}]^{[r]/r} \le 2^{2[r]-1}[E(T_N^0 - S_N^0)^{2[r]} + E(S_N^0 - \hat{S}_N^0)^{2[r]}].$$

Now using Lemmas 3.3 and 3.4 to the terms on the right-hand side, we obtain, after omitting some details of computations, that

$$[E(T_N^0 - \hat{S}_N^0)^{2r}]^{[r]/r} \le \{32e(\Delta + \|\phi_1\|)\}^{2[r]}(2[r])^{2[r]}N^{-[r]}.$$

Thus,

$$E(T_N^0 - \hat{S}_N^0)^{2r}$$
$$\le \{32([r]/r)e(\Delta + \|\phi_1\|)\}^{2r}(2r)^{2r}N^{-r} \le \{64e(\Delta + \|\phi_1\|)\}^{2r}(2r)^{2r}N^{-r}$$

where the last inequality follows from the fact that $1 \le [r]/r \le 2$ for any real $r \ge 1/2$.

4. Proof of Theorem 2.1. By assumption (A) and Lemma 3.5, we have

$$|\hat{\sigma}_N^2 - \tau_N^2| = |2\,\mathrm{Cov}(T_N, \hat{S}_N - T_N) + \mathrm{Var}(\hat{S}_N - T_N)|$$
$$\le 2\tau_N(\mathrm{Var}(\hat{S}_N - T_N))^{1/2} + \mathrm{Var}(\hat{S}_N - T_N)$$

which implies that $\hat{\sigma}_N^2/\tau_N^2 = 1 + O(N^{-1/2})$ and that there is a positive constant σ such that for all $N \ge N_0$,

$$\hat{\sigma}_N^2 \ge \sigma^2. \tag{4.1}$$

Hence, in order to prove Theorem 2.1, it is sufficient to show that

$$P(T_N - ET_N > \hat{\sigma}_N x)[1 - \Phi(x)]^{-1} = 1 + o(1) \tag{4.2}$$

uniformly in the region $0 < x < \rho_N N^{1/6}$.

By standard arguments we have

$$\begin{aligned} &P(T_N^0 > \hat{\sigma}_N x) \le P(\hat{S}_N^0 > (x - N^{-1/6})\hat{\sigma}_N) + P(|T_N^0 - \hat{S}_N^0| > N^{-1/6}\hat{\sigma}_N),\\ &P(T_N^0 > \hat{\sigma}_N x) > P(\hat{S}_N^0 > (x + N^{-1/6})\hat{\sigma}_N) - P(|T_N^0 - \hat{S}_N^0| > N^{-1/6}\hat{\sigma}_N). \end{aligned} \tag{4.3}$$

Using Chebyshev's inequality and applying Lemma 3.5 with $r = \frac{1}{2}\delta N^{1/3}$, $\delta = \sigma(64(\Delta + \|\phi_1\|)e^2)^{-1}$, we get

$$P(|T_N^0 - \hat{S}_N^0| > N^{-1/6}\hat{\sigma}_N) \le E(T_N^0 - \hat{S}_N^0)^{2r}\, N^{r/3}\hat{\sigma}_N^{-2r} \le \exp(-\delta N^{1/3}) \tag{4.4}$$

which implies that uniformly for $0 < x \le \rho_N N^{1/6}$,

$$\begin{aligned} &P(|T_N^0 - \hat{S}_N^0| > N^{-1/6}\hat{\sigma}_N)[1 - \Phi(x)]^{-1}\\ &\qquad \le \exp(-\delta N^{1/3})[1 - \Phi(\rho_N N^{1/6})]^{-1} = o(1) \end{aligned} \tag{4.5}$$

where the last equality follows by Lemma VII.1.2 of Feller (1968).

Now, in view of (4.3) and (4.5), to prove (4.2) it suffices to show that as $N \to \infty$

$$P(\hat{S}_N^0 > x_N \hat{\sigma}_N)[1 - \Phi(x)]^{-1} = 1 + o(1) \tag{4.6}$$

uniformly in $|x_N - x| = N^{-1/6}$, $0 \le x \le \rho_N N^{1/6}$.

Using (2.3) and (3.6) we can write $\hat{S}_N^0 = \sum_{j=1}^N \hat{S}_N^{(j)}$ where

$$\hat{S}_N^{(j)} = c_j[\phi(\rho_{jj}, X_j) - E\phi(\rho_{jj}, X_j)] + (N+1)^{-1} \sum_{k \ne j}^N c_k E\{[u(X_k^g - X_j^g) - G_j(X_k^g)]\phi_1(\rho_{kk}, X_k) \mid X_j\}.$$

Thus $\hat{S}_N^0$ is a sum of independent r.v.'s with means zero and

$$|S_N^{(j)}| \le 2|c_j| \, \|\phi\| + (N+1)^{-1} \sum_{k \ne j}^N |c_k| \, \|\phi_1\| \le (2\|\phi\| + \|\phi_1\|) A_1 N^{-1/3}, \quad 1 \le j \le N,$$

where $\|\phi\| = \sup_{0<s<1} \sup_{-\infty<t<\infty} |\phi(s, t)|$. Furthermore, there is an integer N_0 such that for all $N \ge N_0$ and $0 < x < \rho_N N^{1/6}$, $|x_N - x| \le N^{-1/6}$,

$$0 < (2\|\phi\| + \|\phi_1\|) A_1 N^{-1/3} \hat{\sigma}_N^{-1} x_N \le 1/12 \tag{4.7}$$

in view of (4.1). Thus we can use Theorem 1 of Feller (1943) (cf. also Petrov, 1975, page 253) to obtain that for all $N \ge N_0$,

$$P(\hat{S}_N^0 > x_N \hat{\sigma}_N) = \exp\{-\tfrac{1}{2} x_N^2 Q_N(x_N)\}[1 - \Phi(x_N) + \theta_N \lambda_N \exp(-\tfrac{1}{2} x_N^2)] \tag{4.8}$$

where

$$\lambda_N = (2\|\phi\| + \|\phi_1\|) A_1 N^{-1/3} \hat{\sigma}_N^{-1}, \quad \theta_N < 7.465,$$

$$Q_N(x) = \sum_{j=1}^\infty q_{Nj} x^j, \qquad q_{N1} = 3^{-1} \hat{\sigma}_N^{-3} \sum_{j=1}^N E(\hat{S}_N^{(j)})^3, \tag{4.9}$$

$$|q_{Nj}| < 8^{-1}(12\lambda_N)^j, \qquad j \ge 2.$$

Note that $|x_N| \le \rho_N N^{1/6} + N^{-1/6}$. Since $\rho_N = o(1)$ we have $x_N = o(N^{1/6})$ as $N \to \infty$. Let $K > 0$ be such that

$$|x_N| \le \frac{K \hat{\sigma}_N N^{1/6}}{12 A_1 (2\|\phi\| + \|\phi_1\|)}, \qquad N \ge 1. \tag{4.10}$$

Also it follows by assumption (B) and (4.1) that

$$|q_{N1} x_N| \le \hat{\sigma}_N^{-3} \, 4/3 \, (2\|\phi\|^3 + \|\phi_1\|^3) \sum_{j=1}^N |c_j|^3 x_N \le A_3 N^{-1/3} \tag{4.11}$$

where A_3 is an absolute constant. Hence combining (4.9), (4.10) and (4.11) we obtain that

$$|Q_N(x_N)| \le A_3 N^{-1/3} + \sum_{j=2}^\infty |q_{Nj} x_N^j| \le A_3 N^{-1/3} + 8^{-1} \sum_{j=2}^\infty (K N^{-1/6})^j$$

which implies that as $N \to \infty$

$$x_N^2 Q_N(x_N) = x_N^2 O(N^{-1/3}) = o(N^{1/3}) O(N^{-1/3}) = o(1) \tag{4.12}$$

uniformly in $0 < x \le \rho_N N^{1/6}$. Moreover Lemma VII.1.2 of Feller (1968) ensures that

$$\theta_N \lambda_N \exp(-\tfrac{1}{2} x_N^2)[1 - \Phi(x_N)]^{-1} = O(\lambda_N x_N) = o(1),$$
(4.13)
$$[1 - \Phi(x_N)]^{-1}[1 - \Phi(x)] = 1 + o(1).$$

(4.2) now follows uniformly in $0 < x \leq \rho_N N^{1/6}$ by using (4.8), (4.12) and (4.13).

REMARK 4.1. The above theorem obviously holds for the case of the iid r.v.'s. However, for such a case, and for the statistics (1.4) and (1.5), the theorem holds under somewhat weaker assumptions on the score generating function, if we assume somewhat stronger assumption on the underlying distribution function. For the case of the statistic (1.4), we refer to Kallenberg (1982). For the case of the statistic (1.5), we have the following theorem:

THEOREM 4.1. *Let* X_{Nj}, $1 \leq j \leq N$, $N \geq 1$ *be iid r.v.'s with a continuous c.d.f.* $F_N(x)$ *symmetric about* 0. *Assume the following*:

ASSUMPTION (D). $\sum_{j=1}^N c_{Nj}^2 = 1$, $\max_{1\leq j\leq N} | c_{Nj} | \leq A_1 N^{-1/3}$ where A_1 is an absolute constant.

ASSUMPTION (E). *The score generating function* $\tilde{\phi}$ *(defined in Section* 1) *is not identically zero and satisfies a Lipschitz condition of order* 1 *on* (0, 1), *that is, there exists a constant C such that* $| \tilde{\phi}(t) - \tilde{\phi}(s) | \leq C | t - s |$ *for all* $t, s \in (0, 1)$.
Then, the conclusion of the Theorem 2.1 *holds.*

To prove this theorem we use the following two lemmas and proceed essentially as in the proof of Theorem 2.1. (As before, we suppress the subscript N in c_{Nj}, X_{Nj} etc. whenever it causes no confusion).

LEMMA 4.1. *Let* $Z_1, \cdots, Z_N$ *be random variables such that for any permutation* $(i_1, \cdots, i_N)$ *of* $(1, \cdots, N)$

(4.14)
$$E \prod_{j=1}^N Z_j^{\alpha_j} = E \prod_{j=1}^N Z_{i_j}^{\alpha_j}$$

where α_j*'s are nonnegative integers such that* $\sum_{j=1}^N \alpha_j = 2k$, $k \geq 1$ *integer. Furthermore, assume that*

(4.15)
$$E \prod_{j=1}^N Z_j^{\alpha_j} = 0$$

if at least one of the α_j's *is odd. Then if* $\sum_{j=1}^N c_j^2 = 1$, *for any integer* $k \leq d(\max_{1\leq j\leq N} | c_j |)^{-1}$, $d > 0$

(4.16)
$$E(\textstyle\sum_{j=1}^N c_j Z_j)^{2k} \leq 2^{2k+1}(\max\{1, d\})^{2k} k^k E Z_1^{2k}.$$

Consider now the statistic $S_N = \sum_{j=1}^N c_j \tilde{\phi}(F_N^*(| X_j |))\operatorname{sgn} X_j$ where $F_N^*(x) = P(| X_1 | \leq x)$, $0 \leq x < \infty$.

LEMMA 4.2. *For all real* p, $1 \le p \le N$,

$$E(T_N - S_N)^{2p} \le A_2^{2p}(2p)^{2p}N^{-p}(\max\{1, p \max_{1\le j\le N} |c_j|\})^{2p}$$

where A_2 *is a constant independent of* N *and* p.

PROOF OF LEMMA 4.1. Because of (4.14) and (4.15), the multinomial expansion yields

$$E(\textstyle\sum_{j=1}^N c_j Z_j)^{2k}$$

$$(4.17)\quad = \sum_{\alpha=1}^k \sum_{(k_1,\cdots,k_\alpha)\in A_\alpha} \frac{(2k)!}{\prod_{\nu=1}^\alpha [(2\nu)!]^{k_\nu} \prod_{\nu=1}^\alpha (k_\nu)!} \cdot \sum_{(i_1,\cdots,i_{K_\alpha})\neq} E\prod_{j=1}^{K_1} (c_{i_j}Z_{i_j})^2 \prod_{j=K_1+1}^{K_2} (c_{i_j}Z_{i_j})^4 \cdots \prod_{j=K_{\alpha-1}}^{K_\alpha} (c_{i_j}Z_{i_j})^{2\alpha}$$

where k_j, $1 \le j \le \alpha$ are nonnegative integers, $K_\beta = \sum_{\nu=1}^\beta k_\nu$, $1 \le \beta \le \alpha$, $A_\alpha = \{(k_1, k_2, \cdots, k_\alpha): \sum_{\nu=1}^\alpha \nu k_\nu = k\}$ and $\sum_{(i_1,\cdots,i_{K_\alpha})\neq}$ means that the sum is taken over mutually different indices $1 \le i_1, i_2, \cdots i_{K_\alpha} \le N$.

From (4.14) and generalized Hölder's inequality, it follows that

$$(4.18)\quad \sum_{(i_1,\cdots,i_{K_\alpha})\neq} E\prod_{j=1}^{K_1} (c_{i_j}Z_{i_j})^2 \prod_{j=K_1+1}^{K_2} (c_{i_j}Z_{i_j})^4 \cdots \prod_{j=K_{\alpha-1}}^{K_\alpha} (c_{i_j}Z_{i_j})^{2\alpha} \le \sum_{(i_1,\cdots,i_{K_\alpha})\neq} \pi(k_1, k_2, \cdots, k_\alpha)EZ_1^{2k}$$

where

$$\pi(k_1, k_2, \cdots, k_\alpha) = \prod_{j=1}^{K_1} c_{i_j}^2 \prod_{j=K_1+1}^{K_2} c_{i_j}^4 \cdots \prod_{j=K_{\alpha-1}+1}^{K_\alpha} c_{i_j}^{2\alpha}.$$

Now, if we let $p_1 = \cdots = p_{K_1} = 2$,

$$p_{K_1+1} = \cdots = p_{K_2} = 4, \cdots, p_{K_{\alpha-1}+1} = \cdots = p_{K_\alpha} = 2\alpha,$$

by using the conditions $k \le d(\max_{1\le j\le N} |c_j|)^{-1}$ and $\sum_{j=1}^N c_j^2 = 1$; we get

$$(4.19)\quad \sum_{(i_1,\cdots,i_{k_\alpha})\neq} \pi(k_1, \cdots, k_\alpha) \le \sum_{\ell_1,\cdots,\ell_{K_\alpha}=1}^N c_{\ell_1}^{p_1} \cdots c_{\ell_{K_\alpha}}^{p_{K_\alpha}} \le \prod_{j=1}^{K_\alpha}(\textstyle\sum_{i=1}^N c_i^{p_j}) \le (\max |c_i|)^{2k-2K_\alpha} \le (d/k)^{2k-2K_\alpha}.$$

Thus, relations (4.17)–(4.19) yield the estimate

$$(4.20)\quad E(\textstyle\sum_{j=1}^N c_j Z_j)^{2k} \le \sum_{\alpha=1}^k \sum_{(k_1,\cdots,k_\alpha)\in A_\alpha} \frac{(2k)!}{\prod_{\nu=1}^\alpha [(2\nu)!]^{k_\nu} \prod_{\nu=1}^\alpha (k_\nu)!} \left(\frac{d}{k}\right)^{2k-2K_\alpha} EZ_1^{2k}.$$

Further, by Stirling's formula (see Feller, 1968)

$$(2\pi)^{1/2}n^{n+1/2}e^{-n} < n! \le (2\pi)^{1/2}n^{n+1/2}e^{-n} (1 + (1/4n))$$

we obtain, after some computations, that

$$(4.21)\quad \frac{\prod_{\nu=1}^\alpha (\nu!)^{k_\nu}(2k)!(k - K_\alpha)!}{\prod_{\nu=1}^\alpha [(2\nu)!]^{k_\nu}(k!)^2} \left(\frac{d}{k}\right)^{2k-2K_\alpha} \le 2^{2k+1}(\max\{1, d\})^{2k}.$$

Finally, on the basis of the expansion

$$k^k = \sum_{\alpha=1}^{k} \sum_{(k_1,\cdots,k_\alpha)\in A_\alpha} (k!)^2/\{\prod_{\nu=1}^{\alpha} (\nu!)^{k_\nu}\prod_{\nu=1}^{\alpha} (k_\nu)!(k - K_\alpha)!\}$$

inequalities (4.20) and (4.21) lead to the assertion (4.16) in Lemma 4.1.

PROOF OF LEMMA 4.2. Let $1 \le p \le N$ be any given integer and let $Z_j = \{\tilde{\phi}(R_j^+/(N + 1)) - \tilde{\phi}(F^*(|X_j|))\} \operatorname{sgn} X_j$, $1 \le j \le N$. Then, using Lemma 4.1 with $d = p \max_{1\le j\le N} |c_j|$, it follows that

$$E(T_N - S_N)^{2p} \le 2^{2p+1}(\max\{1, p \max_{1\le j\le N} |c_j|\})^{2p}p^pEZ_1^{2p}. \tag{4.22}$$

Now, using Hölder's inequality and proceeding as in Kallenberg (1982), we obtain the desired result for any real p, $1 \le p \le N$. The details are omitted.

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