

WEAK CONVERGENCE OF GENERALIZED U -STATISTICS¹

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Wichura (1969) studied an invariance principle for partial sums of a multi-dimensional array of independent random variables. It is shown that a similar invariance principle holds for a broad class of generalized U -statistics for which the different terms in the partial sums are not independent. Weak convergence of generalized U -statistics for random sample sizes is also studied. The case of (generalized) von Mises' functional is treated briefly.

1. Introduction. Let $\{X_{ji}, i \geq 1\}$, $j = 1, \dots, c$, be c (≥ 2) independent sequence of independent random vectors (irv), where X_{ji} has a distribution function (df) $F_j(x)$, $x \in R^p$, the p (≥ 1)-dimensional Euclidean space, for $j = 1, \dots, c$. Let $g(X_{ji}, i = 1, \dots, m_j, j = 1, \dots, c)$ be a Borel-measurable kernel of degree $\mathbf{m} = (m_1, \dots, m_c)$, where we assume (without any loss of generality) that g is symmetric in the m_j (≥ 1) arguments (vectors) of the j th set, for $j = 1, \dots, c$. Let $m_0 = m_1 + \dots + m_c$, $\mathbf{F} = (F_1, \dots, F_c)$, and consider a functional of \mathbf{F}

$$(1.1) \quad \theta(\mathbf{F}) = \int_{R^{pm_0}} \dots \int g(x_{11}, \dots, x_{cm_c}) \prod_{j=1}^c \prod_{i=1}^{m_j} dF_j(x_{ji})$$

defined on $\mathcal{F} = \{\mathbf{F}: |\theta(\mathbf{F})| < \infty\}$. $\theta(\mathbf{F})$ is called an *estimable parameter* or a *regular functional* of \mathbf{F} over \mathcal{F} .

For a set of samples of sizes $\mathbf{n} = (n_1, \dots, n_c)$ with $n_j \geq m_j$, $1 \leq j \leq c$, the *generalized U -statistic* for $\theta(\mathbf{F})$ is defined by

$$(1.2) \quad U(\mathbf{n}) = \prod_{j=1}^c \binom{n_j}{m_j}^{-1} \sum_{(\mathbf{n})}^* g(X_{j\alpha}, \alpha = i_{j1}, \dots, i_{jm_j}, 1 \leq j \leq c),$$

where the summation $\sum_{(\mathbf{n})}^*$ extends over all $1 \leq i_{j1} < \dots < i_{jm_j} \leq n_j$, $1 \leq j \leq c$. For various properties of $U(\mathbf{n})$, including its asymptotic normality, we may refer to Fraser (1957) and Puri and Sen (1971), among others. For the asymptotic normality, it is assumed that $\theta(\mathbf{F})$ is *stationary of order zero* [i.e., (2.3) holds] and essentially

$$(1.3) \quad \lim_{n \rightarrow \infty} n_j/n = \lambda_j: 0 < \lambda_j < 1, \quad j = 1, \dots, c,$$

where $n = n_1 + \dots + n_c$. Weak convergence of a stochastic process derived

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from the tail sequence of one-sample U -statistics to a Wiener process has been studied by Loynes (1970), while Miller and Sen (1972) consider a Donsker-type invariance principle for one-sample U -statistics. They show that a process derived from $\{U([k\lambda_1], \dots, [k\lambda_c]), k \leq n\}$ converges weakly to a one-dimensional Wiener process. The more general and natural case of a c -dimensional time-parameter where we use the entire set $\{U(\mathbf{k}): \mathbf{m} \leq \mathbf{k} \leq \mathbf{n}\}$ (or the entire tail set $\{U(\mathbf{k}): \mathbf{k} \geq \mathbf{n}\}$) is considered here, and it is shown that weak convergence to appropriate multi-dimensional Gaussian processes hold under no extra regularity conditions; here $\mathbf{a} \leq \mathbf{b}$ means that $a_i \leq b_i$ for all $1 \leq i \leq c$. It may be noted that for $c \geq 2$, the ordering of \mathbf{n} is not defined, and as a result, the treatment of Loynes (1970) or of Miller and Sen (1972), resting on the reverse martingale property of U -statistics, does not work out. Also, as is usually the case with generalized U -statistics, $U(\mathbf{n})$ in (1.2) involves a set of summands which are not all stochastically independent. Thus, Theorem 2 (or its Corollary 1) of Wichura (1969) does not lead us to the desired result. The task is accomplished here by first extending Theorem 1 of Wichura (1969) to more general summands, and then using a decomposition of $U(\mathbf{n})$ which fits into this extension.

The preliminary notions and the basic theorems are considered in Section 2. The proofs of the theorems are presented in Section 3. The case of generalized von Mises' (1947) functionals is treated in Section 4. In the last section, the case of random sample sizes is also considered. The results are useful in the developing area of sequential inference based on generalized U -statistics.

2. Statement of the main results. For every $d_j: 0 \leq d_j \leq m_j, 1 \leq j \leq c$, let

$$(2.1) \quad g_{d_1 \dots d_c}(x_{j_i}, i = 1, \dots, d_j, 1 \leq j \leq c) \\ = E\{g(x_{j_1}, \dots, x_{j_{d_j}}, X_{j_{d_j+1}}, \dots, X_{j_{m_j}}, 1 \leq j \leq c)\},$$

so that $g_{0 \dots 0} = \theta(\mathbf{F})$ and $g_{m_1 \dots m_c}(\) = g(\)$. Let then

$$(2.2) \quad \zeta_{d_1 \dots d_c}(\mathbf{F}) = E g_{d_1 \dots d_c}^2(X_{j_i}, 1 \leq i \leq d_j, 1 \leq j \leq c) - \theta^2(\mathbf{F}),$$

so that $\zeta_{0 \dots 0}(\mathbf{F}) = 0$. We assume that

(i) $\theta(\mathbf{F})$ is stationary of order 0, i.e.,

$$(2.3) \quad \sigma_j^2 = m_j^2 \zeta_{\delta_{j_1} \dots \delta_{j_c}}(\mathbf{F}) > 0 \quad \text{for every } 1 \leq j \leq c,$$

(where $\delta_{ab} = 1$ or 0 according as $a = b$ or not), and

(ii) g is square integrable, i.e.,

$$(2.4) \quad \zeta_{m_1 \dots m_c}(\mathbf{F}) < \infty \quad \text{i.e.,} \quad g \in L^2.$$

Later on, we shall see that (2.3) may be replaced by $\max_{1 \leq j \leq c} \sigma_j^2 > 0$, and the necessary modifications are trivial. We know that for $n_j \geq m_j, 1 \leq j \leq c$, under (1.3), (2.3) and (2.4).

$$(2.5) \quad r^2(\mathbf{n}) = V(U(\mathbf{n})) = \prod_{j=1}^c \binom{n_j}{m_j}^{-1} \sum_{d_1=0}^{m_1} \dots \sum_{d_c=0}^{m_c} \prod_{j=1}^c \left\{ \binom{m_j}{d_j} \binom{n_j - m_j}{m_j - d_j} \right\} \zeta_{d_1 \dots d_c}(\mathbf{F}) \\ = \sum_{j=1}^c n_j^{-1} \sigma_j^2 + O(n^{-2}), \quad n = n_1 + \dots + n_c.$$

Let now $E_c = [0, 1]^c$ be the c -dimensional unit cube in R^c , $\mathbf{t} = (t_1, \dots, t_c) \in E_c$, and let $[\mathbf{nt}] = ([n_1 t_1], \dots, [n_c t_c])$ where $[s]$ denotes the largest integer $\leq s$. Then, for every $\mathbf{n} (\geq \mathbf{m})$, we define a process $W(\mathbf{n}) = \{W(\mathbf{t}; \mathbf{n}) : \mathbf{t} \in E_c\}$ by letting

$$(2.6) \quad W(\mathbf{t}, \mathbf{n}) = \phi([\mathbf{nt}]; n)[U([\mathbf{nt}]) - \theta(\mathbf{F})], \\ \forall [\mathbf{nt}] \geq \mathbf{m}, \text{ and } 0, \text{ otherwise,}$$

where for $\mathbf{k} = (k_1, \dots, k_c)$, $k_j > 0$, $j = 1, \dots, c$,

$$(2.7) \quad \phi(\mathbf{k}, n) = n^{-\frac{1}{2}} (\sum_{j=1}^c \sigma_j \lambda_j^{\frac{1}{2}}) (\sum_{j=1}^c \sigma_j \lambda_j^{\frac{1}{2}} k_j^{-1})^{-1},$$

so that $\phi(\mathbf{k}, n)$ is $n^{-\frac{1}{2}}$ times a harmonic mean of k_1, \dots, k_c . Consider now c independent copies of a standard Brownian motion on $[0, 1]$ and denote these by $W_j = \{W_j(t) : 0 \leq t \leq 1\}$, $j = 1, \dots, c$. Finally, the space $D_c = D\{[0, 1]^c\}$ of all real functions on E_c with no discontinuities of the second kind and associated (extended) Skorokhod J_1 -topology are defined as in Neuhaus (1971). Then, we have the following.

THEOREM 2.1. *Under (1.3), (2.3) and (2.4), $W(\mathbf{n})$ converges in law in the extended Skorokhod J_1 -topology on D_c to a Gaussian function $W = \{W(\mathbf{t}) : \mathbf{t} \in E_c\}$, where*

$$(2.8) \quad W(\mathbf{t}) = (\sum_{j=1}^c \sigma_j \lambda_j^{\frac{1}{2}}) (\sum_{j=1}^c \sigma_j \lambda_j^{-\frac{1}{2}} t_j^{-1})^{-1} [\sum_{j=1}^c \sigma_j \lambda_j^{-\frac{1}{2}} t_j^{-1} W_j(t_j)], \mathbf{t} > \mathbf{0}, \\ = 0, \text{ with probability } 1, \text{ if } t_j = 0 \text{ for some } j: 1 \leq j \leq c.$$

We define a process $W^*(\mathbf{n}) = \{W^*(\mathbf{t}; \mathbf{n}) : \mathbf{t} \in E_c\}$ as follows. Considering the tail set $\{U(\mathbf{k}) : \mathbf{k} \geq \mathbf{n}\}$, let

$$(2.9) \quad W^*(\mathbf{t}; \mathbf{n}) = r^{-1}(\mathbf{n})[U([\mathbf{n}/\mathbf{t}]) - \theta(\mathbf{F})], \quad \mathbf{t} \in E_c,$$

where $[\mathbf{n}/\mathbf{t}] = ([n_1/t_1], \dots, [n_c/t_c])$. Let then

$$(2.10) \quad \mathbf{w} = (w_1, \dots, w_c)'; w_j = \sigma_j \lambda_j^{-\frac{1}{2}} (\sum_{j=1}^c \sigma_j^2 / \lambda_j)^{-\frac{1}{2}}, \quad 1 \leq j \leq c;$$

$$(2.11) \quad \mathbf{w}(\mathbf{t}) = (W_1(t_1), \dots, W_c(t_c))', \quad \mathbf{t} \in E_c,$$

where the $W_j(t)$ are defined earlier, and let

$$(2.12) \quad W^* = \{W^*(\mathbf{t}) : \mathbf{t} \in E_c\}; W^*(\mathbf{t}) = \mathbf{w}' \mathbf{W}(\mathbf{t}), \quad \mathbf{t} \in E_c.$$

THEOREM 2.2. *Under (1.3), (2.3) and (2.4), as $n \rightarrow \infty$,*

$$(2.13) \quad W^*(\mathbf{n}) \rightarrow_{\mathcal{D}} W^*, \text{ in the Skorokhod } J_1\text{-topology on } D_c.$$

Theorems 2.1 and 2.2 provide multi-sample extensions of the weak convergence results of Miller and Sen (1972) and Loynes (1970), respectively. Related results on von Mises' (1947) functionals are considered in Section 4.

3. Proofs of the theorems. For simplicity of the proofs, we explicitly consider the case of $c = 2$; an essentially same but more laborious proof holds for general $c (\geq 2)$. First, we consider certain basic lemmas needed in the subsequent steps of the proof.

Let $\mathcal{B}_n^{(j)}$ be the σ -field generated by $\{X_{j1}, \dots, X_{jn}\}$ for $n \geq 1, 1 \leq j \leq c$; $\mathcal{B}_{n_1 n_2}^{(12)}$ denote the product σ -field $\mathcal{B}_{n_1}^{(1)} \times \mathcal{B}_{n_2}^{(2)}$ for $n_1 \geq 1, n_2 \geq 1$. Further, let $S_{ii'} = S(X_{11}, \dots, X_{1i}, X_{21}, \dots, X_{2i'})$, for $i, i' \geq 1$, be a sequence of random variables such that

$$(3.1) \quad E(S_{ij} | \mathcal{B}_{i'j}^{(12)}) = S_{i'j} \quad \text{a.e. (almost everywhere)}$$

for every $i \geq i' \geq 1$ and $j \geq 1$, and for every $j \geq j' \geq 1, n_1 \geq 1$,

$$(3.2) \quad E(\mathbf{S}_{n_1}^{(j)} | \mathcal{B}_{n_1 j'}^{(12)}) = \mathbf{S}_{n_1}^{(j')} \quad \text{a.e.,}$$

where $\mathbf{S}_k^{(j)} = (S_{1j}, \dots, S_{kj})'$ for $k \geq 1, j \geq 1$. Finally, assume that for every $i \geq 1, j \geq 1$,

$$(3.3) \quad E(S_{ij}) = 0, \quad \sigma_{ij}^2 = E(S_{ij}^2) < \infty.$$

LEMMA 3.1. *Under (3.1), (3.2) and (3.3), for every $n_1 \geq 1, n_2 \geq 1$,*

$$(3.4) \quad E[(\max_{1 \leq i \leq n_1} \max_{1 \leq j \leq n_2} |S_{ij}|)^2] \leq 16\sigma_{n_1 n_2}^2.$$

Since Doob's ((1953) page 317) inequality holds for nonnegative submartingales, the proof of the lemma follows precisely on the same line as in the proof of Theorem 1 of Wichura (1969), and hence, the details are omitted. The extension of (3.4) for a general $c (\geq 2)$ is immediate; we need to replace 16 by 4^c and $\sigma_{n_1 n_2}^2$ by $E[S^2(X_{11}, \dots, X_{1n_1}, \dots, X_{c1}, \dots, X_{cn_c})]$.

Consider now a two-sample U -statistic $U_{n_1 n_2}$ and denote by $r^2(n_1, n_2) = \text{Var}(U_{n_1 n_2})$; note that by (2.5), $r^2(n_1, n_2)$ is non-increasing in each of n_1 and n_2 .

LEMMA 3.2. *For every $N_j \geq n_j \geq m_j (> 1), j = 1, 2$,*

$$(3.5) \quad \begin{aligned} E[(\max_{n_1 \leq k \leq N_1} \max_{n_2 \leq q \leq N_2} r^{-1}(n_1, n_2) |U_{kq} - U_{kN_2} - U_{N_1q} + U_{N_1 N_2}|)^2] \\ \leq 16r^{-2}(n_1, n_2)[r^2(n_1, n_2) - r^2(n_1, N_2) - r^2(N_1, n_2) + r^2(N_1, N_2)] \\ \leq 16, \quad \text{uniformly in } N_1(\geq n_1) \text{ and } N_2(\geq n_2). \end{aligned}$$

PROOF. For every $r \geq 1, s \geq 1$, we let

$$(3.6) \quad h_{rs}^{(N_1 N_2)} = U_{N_1-r+1N_2-s+1} - U_{N_1-rN_2-s+1} - U_{N_1-r+1N_2-s} + U_{N_1-rN_2-s},$$

so that for every $1 \leq i \leq N_1 - n_1, 1 \leq j \leq N_2 - n_2$,

$$(3.7) \quad \begin{aligned} S_{ij} = \sum_{r=1}^i \sum_{s=1}^j h_{rs}^{(N_1 N_2)} = U_{N_1-i+1N_2-j+1} - U_{N_1-i+1N_2} \\ - U_{N_1 N_2-j+1} + U_{N_1 N_2}. \end{aligned}$$

Let, now $\mathcal{C}_n^{(j)}$ be the σ -field generated by the unordered collection $\{X_{j1}, \dots, X_{jn}\}$ and by $X_{jn+1}, X_{jn+2}, \dots, j = 1, 2$, and $\mathcal{C}_{n_1 n_2}^{(12)}$ be the product σ -field $\mathcal{C}_{n_1}^{(1)} \times \mathcal{C}_{n_2}^{(2)}$ for $n_1 \geq m_1, n_2 \geq m_2$. It follows by standard arguments that $E(U_{kq} | \mathcal{C}_{k'q'}^{(1)}) = U_{k'q'}$, a.e. for every $k \leq k'$ and $q \leq q'$. Consequently, it follows by some routine steps that for every $1 \leq i \leq N_1 - n_1, 1 \leq j \leq N_2 - n_2$,

$$(3.8) \quad E(S_{ij} | \mathcal{C}_{i'j}^*) = S_{i'j} \quad \text{a.e.,} \quad i' \leq i,$$

$$(3.9) \quad E(\mathbf{S}_{N_1-n_1}^{(j)} | \mathcal{C}_{N_1-n_1, j'}^*) = \mathbf{S}_{N_1-n_1}^{(j')} \quad \text{a.e.,} \quad j \geq j',$$

where $\mathcal{C}_{kq}^* = \mathcal{C}_{N_1-k, N_2-q}^{(12)}$, $1 \leq k \leq N_1 - n_1$, $1 \leq q \leq N_2 - n_2$, and $\mathbf{S}_k^{(j)} = (S_{1j}, \dots, S_{kj})'$, for $k \geq 1$, $j \geq 1$. Further, by (3.7), $E(S_{ij}) = 0$ for all $i \geq 1$, $j \geq 1$. Finally, \mathcal{C}_{kq}^* is \uparrow in k and q . Consequently, the same proof as in Lemma 3.1 holds, and (3.5) follows by noting that by (3.7),

$$(3.10) \quad E(S_{N_1-n_1, N_2-n_2}^2) = r^2(n_1, n_2) - r^2(n_1, N_2) - r^2(N_1, n_2) + r^2(N_1, N_2) \\ \leq r^2(n_1, n_2) \quad \text{for every } N_1 \geq n_1, N_2 \geq n_2,$$

as $r^2(i, j)$ is \downarrow in i and j . \square

We further note that $\{U_{kN_2}, \mathcal{C}_{kN_2}; n_1 \leq k \leq N_1\}$ has the reverse martingale property, so that by reversing the index set, we obtain that

$$(3.11) \quad E[(\max_{n_1 \leq k \leq N_1} r^{-1}(n_1, N_2) |U_{kN_2} - U_{N_1N_2}|)^2] \\ \leq 4r^{-2}(n_1, N_2)[r^2(n_1, N_2) - r^2(N_1, N_2)] \leq 4,$$

uniformly in $N_1 \geq n_1$, $N_2 \geq n_2$. Similarly,

$$(3.12) \quad E[(\max_{n_2 \leq q \leq N_2} r^{-1}(N_1, n_2) |U_{N_1q} - U_{N_1N_2}|)^2] \\ \leq 4r^{-2}(N_1, n_2)[r^2(N_1, n_2) - r^2(N_1, N_2)] \leq 4,$$

uniformly in $N_1 \geq n_1$, $N_2 \geq n_2$. Finally,

$$(3.13) \quad E[|U_{N_1N_2} - \theta(\mathbf{F})|^2] = r^2(N_1, N_2) \leq r^2(n_1, n_2),$$

uniformly in $N_1 \geq n_1$, $N_2 \geq n_2$. Hence, by Lemma 3.2, (3.11), (3.12), (3.13), the c_r -inequality and the Chebychev inequality, we obtain that for every $\varepsilon > 0$, there exists a positive $K_\varepsilon (< \infty)$, such that for every $N_1 \geq n_1$, $N_2 \geq n_2$,

$$(3.14) \quad P\{\max_{n_1 \leq k \leq N_1} \max_{n_2 \leq q \leq N_2} |U_{kq} - \theta(\mathbf{F})| > r(n_1, n_2)K_\varepsilon\} < \varepsilon,$$

and hence,

$$(3.15) \quad P\{\sup_{\mathbf{k} \geq \mathbf{n}} |U_{\mathbf{k}} - \theta(\mathbf{F})| > r(\mathbf{n})K_\varepsilon\} < \varepsilon.$$

We now consider a typical decomposition for $U(\mathbf{n})$. Let us write $U_{00}^*(\mathbf{n}) = \theta(\mathbf{F})$ and for $\mathbf{k} \geq \mathbf{0}$,

$$(3.16) \quad U_{\mathbf{k}}^*(\mathbf{n}) = \sum_{d_1=0}^{k_1} \sum_{d_2=0}^{k_2} (-1)^{d_1+d_2} \binom{k_1}{d_1} \binom{k_2}{d_2} U_{d_1d_2}(\mathbf{n}); \quad \mathbf{0} \leq \mathbf{k} \leq \mathbf{m},$$

where for $0 \leq d_1 \leq m_1$ and $0 \leq d_2 \leq m_2$, we have

$$(3.17) \quad U_{d_1d_2}(\mathbf{n}) = \prod_{j=1}^2 \binom{n_j}{d_j}^{-1} \sum_{(\mathbf{n})}^* g_{d_1d_2}(X_{j\alpha_{ji}}, 1 \leq i \leq d_j, 1 \leq j \leq 2),$$

where the summation $\sum_{(\mathbf{n})}^*$ extends over all $1 \leq \alpha_{ji} < \dots < \alpha_{jd_j} \leq m_j$, $j = 1, 2$. Then, by (1.2), (3.16), (3.17) and a few routine steps, we obtain that

$$(3.18) \quad U(\mathbf{n}) = \sum_{k_1=0}^{m_1} \sum_{k_2=0}^{m_2} \binom{m_1}{k_1} \binom{m_2}{k_2} U_{\mathbf{k}}^*(\mathbf{n}),$$

where each $U_{\mathbf{k}}^*(\mathbf{n})$ ($\mathbf{0} \leq \mathbf{k} \leq \mathbf{m}$) is a generalized U -statistic. A few readjustments lead us to

$$(3.19) \quad U(\mathbf{n}) = \theta(\mathbf{F}) + U_1(n_1) + U_2(n_2) + \sum_{d_1=1}^{m_1} \sum_{d_2=1}^{m_2} \binom{m_1}{d_1} \binom{m_2}{d_2} U_{\mathbf{d}}^*(\mathbf{n}),$$

where

$$(3.20) \quad U_1(n_1) = \binom{n_1}{m_1}^{-1} \sum_{(n_1)}^* [g_{m_1 0}(X_{1i_1}, \dots, X_{1i_{m_1}}) - \theta(\mathbf{F})],$$

$$(3.21) \quad U_2(n_2) = \binom{n_2}{m_2}^{-1} \sum_{(n_2)}^* [g_{0m_2}(X_{2i_1}, \dots, X_{2i_{m_2}}) - \theta(\mathbf{F})],$$

and the summation $\sum_{(n_j)}^*$ extends over all $1 \leq i_1 < \dots < i_{m_j} \leq n_j, j = 1, 2$.

We first consider the proof of Theorem 2.2 which is relatively simpler. We write [by (3.19)]

$$(3.22) \quad \begin{aligned} U_3^*(\mathbf{n}) &= [U(\mathbf{n}) - \theta(\mathbf{F})] - U_1(n_1) - U_2(n_2) \\ &= \sum_{d_1=1}^{m_1} \sum_{d_2=1}^{m_2} \binom{m_1}{d_1} \binom{m_2}{d_2} U_{\mathbf{d}}^*(\mathbf{n}), \end{aligned}$$

where $U_3^*(\mathbf{n})$ is a generalized U -statistic for which

$$(3.23) \quad \begin{aligned} [r^*(\mathbf{n})]^2 &= E[U_3^*(\mathbf{n})]^2 = r^2(\mathbf{n}) - E[U_1(n_1)]^2 - E[U_2(n_2)]^2 \\ &= \binom{n_1}{m_1}^{-1} \binom{n_2}{m_2}^{-1} \sum_{d_1=1}^{m_1} \sum_{d_2=1}^{m_2} \binom{m_1}{d_1} \binom{m_2}{d_2} \binom{n_1-m_1}{m_1-d_1} \binom{n_2-m_2}{m_2-d_2} \zeta_{d_1 d_2}(\mathbf{F}) \\ &= n_1^{-1} n_2^{-1} m_1^2 m_2^2 \zeta_{11}(\mathbf{F}) + O(n^{-3}). \end{aligned}$$

Thus, from (2.5) and (3.23), we note that under (1.3),

$$(3.24) \quad r^*(\mathbf{n})/r(\mathbf{n}) = O(n^{-\frac{1}{2}}),$$

and hence, using (3.15) for $\{U_3^*(\mathbf{k}), \mathbf{k} \geq \mathbf{n}\}$ and (3.24), we conclude that for every $\varepsilon > 0$ and $\eta > 0$, there exists an $n_0 = n_0(\varepsilon, \eta)$, such that for $n \geq n_0$,

$$(3.25) \quad P\{\sup_{\mathbf{k} \geq \mathbf{n}} |U_3^*(\mathbf{k})| > \eta \cdot r(\mathbf{n})\} < \varepsilon.$$

Therefore, under (1.3), (2.3) and (2.4), as $n \rightarrow \infty$,

$$(3.26) \quad \sup_{\mathbf{k} \geq \mathbf{n}} r^{-1}(\mathbf{n}) |U(\mathbf{k}) - \theta(\mathbf{F}) - U_1(k_1) - U_2(k_2)| \rightarrow_P 0.$$

Let us now define for each $j (= 1, 2)$,

$$(3.27) \quad W_j^*(t, n_j) = \sigma_j^{-1} n_j^{\frac{1}{2}} U_j([n_j/t]), \quad 0 < t \leq 1, \quad \text{and } 0 \text{ for } t = 0,$$

$$(3.28) \quad w_j(\mathbf{n}) = r^{-1}(\mathbf{n}) \sigma_j n_j^{-\frac{1}{2}}.$$

Then, from (2.9), (3.26), (3.27) and (3.28), we obtain that

$$(3.29) \quad \sup_{t \in E_2} |W^*(\mathbf{t}; \mathbf{n}) - \sum_{j=1}^2 w_j(\mathbf{n}) W_j^*(t_j, n_j)| \rightarrow_P 0 \quad \text{as } n \rightarrow \infty.$$

Now, by (1.3), (2.5), (2.10) and (3.28), we obtain that

$$(3.30) \quad \lim_{n \rightarrow \infty} w_j(\mathbf{n}) = w_j \quad \text{for } j = 1, 2.$$

Also, $W_j^*(n_j) = \{W_j^*(t_j, n_j), 0 \leq t \leq 1\}, j = 1, 2$, are stochastically independent, and by the results of Loynes (1970), as $n \rightarrow \infty$,

$$(3.31) \quad W_j^*(n_j) \rightarrow_{\mathcal{D}} W_j = \{W_j(t) : 0 \leq t \leq 1\}, \quad j = 1, 2$$

in the Skorokhod J_1 -topology on $D[0, 1]$, where W_1 and W_2 are independent copies of a standard Brownian motion. (2.13) follows from (3.27) through (3.31). \square

To prove Theorem 2.1, we note on using (3.23) and some standard inequalities among the $\{\zeta_{\mathbf{d}}(\mathbf{F}) : \mathbf{0} \leq \mathbf{d} \leq \mathbf{m}\}$ that $r^*(\mathbf{n}) \leq C(\mathbf{F})(n_1 n_2)^{-\frac{1}{2}}, \forall \mathbf{n} \geq \mathbf{m}$, where $C(\mathbf{F}) < \infty$ (whenever (2.4) holds) and it does not depend on \mathbf{n} . First, we show that under (1.3), (2.3) and (2.4) as $n \rightarrow \infty$,

$$(3.32) \quad \max_{\mathbf{m} \leq \mathbf{k} \leq \mathbf{n}} \phi(\mathbf{k}, n) |U_3^*(\mathbf{k})| \rightarrow_p 0.$$

For this, we partition the set $A(\mathbf{n}) = \{\mathbf{k} : \mathbf{m} \leq \mathbf{k} \leq \mathbf{n}\}$ into three disjoint subsets: $A_1(\mathbf{n}) = \{\mathbf{k} \in A(\mathbf{n}) : \min_{1 \leq j \leq 2} k_j \leq \varepsilon_1 n^{\frac{1}{2}}\}$, $A_2(\mathbf{n}) = \{\mathbf{k} \in A(\mathbf{n}) : \varepsilon_1 n^{\frac{1}{2}} < \min_{1 \leq j \leq 2} k_j \leq [n^{\frac{1}{2}}]\}$, and $A_3(\mathbf{n}) = \{\mathbf{k} \in A(\mathbf{n}) : \min_{1 \leq j \leq 2} k_j > [n^{\frac{1}{2}}]\}$, where $\varepsilon_1 (> 0)$ is an arbitrarily small number. The proof of (3.32) rests on the fact that $[\max_{\mathbf{k}} \phi(\mathbf{k}, n) \cdot \min_{\mathbf{k}} r(\mathbf{k})]$ for $\mathbf{k} \in A_i(\mathbf{n}), i = 1, 2, 3$, all converge to zero, as $n \rightarrow \infty$. The proof follows by using Lemma 3.2 and a few routine steps with are omitted for intended brevity.

On $D[0, 1]$, we now consider independent processes $W_j(n_j) = (W_j(t, n_j) : 0 \leq t \leq 1), j = 1, 2$, where for $0 \leq t \leq 1$,

$$(3.33) \quad \begin{aligned} W_j(t, n_j) &= ([n_j t] / \sigma_j n_j^{\frac{1}{2}}) U_j([n_j t]), & m_j \leq [n_j t] \leq n_j, \\ &= 0, & t \leq (m_j - 1) / n_j; & j = 1, 2. \end{aligned}$$

Also, let for every $\mathbf{t} > \mathbf{0}$.

$$(3.34) \quad \begin{aligned} w_j(\mathbf{t}, \mathbf{n}) &= \phi([\mathbf{t}\mathbf{n}]; n) \{ \sigma_j n_j^{\frac{1}{2}} / [n_j t_j] \} \\ &= n^{-\frac{1}{2}} (\sum_{j=1}^2 \sigma_j \lambda_j^{\frac{1}{2}}) (\sum_{j=1}^2 \sigma_j \lambda_j^{\frac{1}{2}} / [n_j t_j])^{-1} (\sigma_j n_j^{\frac{1}{2}} / [n_j t_j]) \\ &= (\sum_{j=1}^2 \sigma_j \lambda_j^{\frac{1}{2}}) (\sum_{j=1}^2 \sigma_j \lambda_j^{\frac{1}{2}} / [n_j t_j])^{-1} (\lambda_j^{\frac{1}{2}} \sigma_j / [n_j t_j]) [1 + o(1)]. \end{aligned}$$

Then, for every $n_j t_j \geq 1, 1 \leq j \leq 2$, $w_j(\mathbf{t}, \mathbf{n})$ is positive and is bounded from above by $(\sum_{j=1}^2 \sigma_j \lambda_j^{\frac{1}{2}}) (n_j / n \lambda_j)^{\frac{1}{2}} = (\sum_{j=1}^2 \sigma_j \lambda_j^{\frac{1}{2}}) [1 + o(1)], j = 1, 2$. Also, for every $\mathbf{t} > \mathbf{0}$,

$$(3.35) \quad \begin{aligned} \lim_{n \rightarrow \infty} w_j(\mathbf{t}, \mathbf{n}) &= (\sum_{j=1}^2 \sigma_j \lambda_j^{\frac{1}{2}}) (\sum_{j=1}^2 \sigma_j \lambda_j^{\frac{1}{2}} t_j^{-1})^{-1} (\sigma_j \lambda_j^{-\frac{1}{2}} t_j^{-1}), & j = 1, 2. \end{aligned}$$

Finally, it follows from Miller and Sen (1972) that $W_1(n_1)$ and $W_2(n_2)$ are stochastically independent, each converging in law to a standard Brownian motion. Consequently, for every $\varepsilon > 0$ and $\eta > 0$, there exist a $\delta > 0$ and an n_0 , such that for $n \geq n_0$,

$$(3.36) \quad P\{\sup_{0 \leq t \leq \delta} |W_j(t; n_j)| > \varepsilon\} < \frac{1}{2}\eta, \quad j = 1, 2,$$

$$(3.37) \quad \sup_{0 \leq t \leq 1} |W_j(t; n_j)| = O_p(1), \quad j = 1, 2.$$

Thus, (2.8) follows from (2.6), (2.7), (3.22) and (3.32) through (3.37). \square

4. Weak convergence of von Mises' functionals. We define the empirical dfs by

$$(4.1) \quad F_j(x, n_j) = n_j^{-1} \sum_{i=1}^{n_j} c(x - X_{ji}), \quad x \in R^p, n_j \geq 1, \text{ for } j = 1, \dots, c,$$

where $c(u) = 1$ or 0 according as all the p components of u are nonnegative or

at least one negative; we let $F(\cdot, \mathbf{n}) = (F_1(\cdot, n_1), \dots, F_c(\cdot, n_c))$. Then, the von Mises (1947) functional corresponding to (1.1) is

$$(4.2) \quad \theta(\mathbf{F}(\cdot, \mathbf{n})) = \int_{R^{pm_0}} \dots \int g(x_{11}, \dots, x_{cm_0}) \prod_{j=1}^c \prod_{i=1}^{m_j} dF_j(x_{ji}; n_j).$$

Here, we define a process $\tilde{W}(\mathbf{n}) = \{\tilde{W}(\mathbf{t}; \mathbf{n}); \mathbf{t} \in E_c\}$ as in (2.6), where we replace $\{U(\mathbf{k}); \mathbf{k} \in A(\mathbf{n})\}$ by $\{\theta(\mathbf{F}(\cdot, \mathbf{k})); \mathbf{k} \in A(\mathbf{n})\}$. Similarly, on replacing $\{U(\mathbf{k}); \mathbf{k} \geq \mathbf{n}\}$ by $\{\theta(\mathbf{F}(\cdot, \mathbf{k}); \mathbf{k} \geq \mathbf{n})\}$ in (2.9), we define $\tilde{W}^*(\mathbf{n}) = \{\tilde{W}^*(\mathbf{t}; \mathbf{n}); \mathbf{t} \in E_c\}$. Finally we strengthen (2.4) to

$$(4.3) \quad \zeta^*(\mathbf{F}) = \max_{1 \leq j \leq c} \max_{1 \leq \alpha_{j1} \leq \dots \leq \alpha_{jm_j} \leq m_j} E\{g^2(X_{1\alpha_{j1}}, \dots, X_{c\alpha_{cm_0}})\} < \infty.$$

THEOREM 4.1. *Under (1.3), (2.3) and (4.3), as $n \rightarrow \infty$, $\tilde{W}(\mathbf{n})$ and $\tilde{W}^*(\mathbf{n})$ converge in law in the extended Skorokhod J_1 -topology on D_c to W and W^* , respectively, which are defined in (2.8) and (2.12).*

PROOF. Again, we consider the case of $c = 2$, and for $\mathbf{0} \leq \mathbf{d} \leq \mathbf{m}$, define

$$(4.4) \quad V_{\mathbf{d}}^*(\mathbf{n}) = \int_{R^{pd_0}} \dots \int g_{\mathbf{d}}(x_{11}, \dots, x_{1d_1}, x_{21}, \dots, x_{2d_2}) \times \prod_{j=1}^2 \prod_{i=1}^{d_j} d[F_j(x_{ji}, n_j) - F_j(x_{ji})],$$

where $d_0 = d_1 + d_2$. Then, we may write

$$(4.5) \quad \begin{aligned} \theta(\mathbf{F}(\cdot, \mathbf{n})) &= \sum_{\mathbf{d}=\mathbf{0}}^{\mathbf{m}} \binom{m_1}{d_1} \binom{m_2}{d_2} V_{\mathbf{d}}(\mathbf{n}) \\ &= \theta(\mathbf{F}) + V_1(n_1) + V_2(n_2) + \sum_{\mathbf{d}=\mathbf{1}}^{\mathbf{m}} \binom{m_1}{d_1} \binom{m_2}{d_2} V_{\mathbf{d}}^*(\mathbf{n}), \end{aligned}$$

where

$$(4.6) \quad V_1(n_1) = \int_{R^{pm_1}} \dots \int g_{m_1 0}(x_1, \dots, x_{m_1}) \prod_{i=1}^{m_1} dF_1(x_i, n_1) - \theta(\mathbf{F});$$

$$(4.7) \quad V_2(n_2) = \int_{R^{pm_2}} \dots \int g_{0m_2}(x_1, \dots, x_{m_2}) \prod_{i=1}^{m_2} dF_2(x_i, n_2) - \theta(\mathbf{F}).$$

Now, proceeding as in the proof of Lemma 2.6 of Miller and Sen (1972), we have for every $k \geq m_j$,

$$(4.8) \quad E[V_j(k) - U_j(k)]^2 \leq C(\mathbf{F})k^{-3}, \quad \text{for } j = 1, 2,$$

where, under (4.3), $C(\mathbf{F}) < \infty$, and $U_j(n_j)$, $j = 1, 2$, are defined in (3.20) and (3.21). Since, $\phi(\mathbf{k}, n) \leq n^{-1} \delta^{-1} k_j$ for $j = 1, 2$, by (4.8), for each $j (= 1, 2)$,

$$(4.9) \quad \begin{aligned} P(\max_{\mathbf{m} \leq \mathbf{k} \leq \mathbf{n}} \phi(\mathbf{k}, n) |U_j(k_j) - V_j(k_j)| > \varepsilon) \\ \leq \sum_{k=m_j}^{n_j} n^{-1} \delta^{-2} k^2 E[U_j(k) - V_j(\mathbf{k})]^2 \varepsilon^{-2} \\ \leq C(\mathbf{F}) n^{-1} \delta^{-2} \varepsilon^{-2} \sum_{k=m_j}^{n_j} k^{-1} = C(\mathbf{F})(\delta \varepsilon)^{-2} \cdot O(n^{-1} \log n), \end{aligned}$$

so that for every $\varepsilon > 0$, the right-hand side of (4.9) converges to 0 as $n \rightarrow \infty$. Similarly, on noting that $r^2(\mathbf{n})$, defined by (2.5) is bounded below by $\min_{1 \leq j \leq c} (\sigma_j^2/n_j)$, we obtain that as $n \rightarrow \infty$, under (1.3), for every $\varepsilon > 0$, $j = 1, 2$,

$$(4.10) \quad \begin{aligned} P(\max_{\mathbf{k} \geq \mathbf{n}} r^{-1}(\mathbf{n}) |U_j(k_j) - V_j(k_j)| > \varepsilon) \\ \leq \varepsilon^{-2} r^{-2}(\mathbf{n}) \sum_{k=n_j}^{\infty} E[|U_j(k) - V_j(k)|^2] \\ \leq \varepsilon^{-2} r^{-2}(\mathbf{n}) C(\mathbf{F}) \cdot \sum_{k=n_j}^{\infty} k^{-3} \\ = C(\mathbf{F}) \varepsilon^{-2} [O(n^{-1})] \rightarrow 0 \quad \text{as } n \rightarrow \infty. \end{aligned}$$

Consequently, to prove the theorem, all we need to show is this under (1.3), (2.3) and (4.3), for every $\mathbf{d} > \mathbf{1}$,

$$(4.11) \quad P\{\max_{\mathbf{m} \leq \mathbf{k} \leq \mathbf{n}} \psi(\mathbf{k}, n) |U_{\mathbf{d}}^*(\mathbf{k}) - V_{\mathbf{d}}^*(\mathbf{k})| > \varepsilon\} < \eta, \quad .$$

$$(4.12) \quad P\{\max_{\mathbf{k} \geq \mathbf{n}} r^{-1}(\mathbf{n}) |U_{\mathbf{d}}^*(\mathbf{k}) - V_{\mathbf{d}}^*(\mathbf{k})| > \varepsilon\} < \eta,$$

where both $\varepsilon (> 0)$ and $\eta (> 0)$ are arbitrarily small, and n is chosen adequately large. Note that $V_{11}^*(\mathbf{k}) = U_{11}^*(\mathbf{k})$ for all $\mathbf{k} > \mathbf{m}$. Also, by extending the proof of Lemma 2.6 of Miller and Sen (1972), we have for every $\mathbf{d} \geq \mathbf{1}$ [under (4.3)],

$$(4.13) \quad E\{[U_{\mathbf{d}}^*(\mathbf{k}) - V_{\mathbf{d}}^*(\mathbf{k})]^2\} \leq C(\mathbf{F})k_1^{-d_1}k_2^{-d_2}\{k_1^{-2} + k_2^{-2}\},$$

for $\mathbf{k} \geq \mathbf{m}$. Consequently, if $d_1 \geq 2$, $d_2 \geq 2$, the proof for (4.11) and (4.12) follow trivially by using (4.13), the Chebychev and the Bonferroni inequalities. Thus, we need to consider the case where $\min_{1 \leq j \leq 2} d_j = 1$, but $d_1 + d_2 \geq 3$. We consider explicitly the case of $\mathbf{d} = (1, 2)$; the case of $(2, 1)$ follows similarly, while for any other $\mathbf{d} : 1 = \min(d_1, d_2) < \max(d_1, d_2) (\geq 2)$, a similar but more laborious proof holds.

By direct evaluation from (3.16) and (4.4), we have

$$(4.14) \quad V_{12}^*(\mathbf{k}) - U_{12}^*(\mathbf{k}) = \frac{1}{k_2} U_{12}^*(\mathbf{k}) + \frac{1}{k_2 - 1} U_{12}^{**}(\mathbf{k}),$$

for every $\mathbf{k} \geq \mathbf{m}$, where for $m_2 \geq 2$,

$$(4.15) \quad U_{12}^{**}(\mathbf{k}) = \frac{1}{k_1 k_2} \sum_{i_1=1}^{k_1} \sum_{i_2=1}^{k_2} \int_{R^{3p}} \cdots \int g_{12}(x_1, y_1, y_2) \\ \times d[c(x_1 - X_{1i_1}) - F_1(x_1)] \\ \times d[c(y_1 - X_{2i_2}) - F_2(y_1)] d[c(y_2 - X_{2i_2}) - F_2(y_2)].$$

Consequently, to prove (4.11) and (4.12) for $\mathbf{d} = (1, 2)$, it suffices to show that as $n \rightarrow \infty$,

$$(4.16) \quad \max_{\mathbf{m} \leq \mathbf{k} \leq \mathbf{n}} |U_{12}^*(\mathbf{k})| = O_p(1), \quad \max_{\mathbf{m} \leq \mathbf{k} \leq \mathbf{n}} |U_{12}^{**}(\mathbf{k})| = O_p(1);$$

$$(4.17) \quad \sup_{\mathbf{k} \geq \mathbf{n}} |U_{12}^*(\mathbf{k})| = O_p(1), \quad \sup_{\mathbf{k} \geq \mathbf{n}} |U_{12}^{**}(\mathbf{k})| = O_p(1);$$

note that $\psi(\mathbf{k}, n)/k_2 = O(n^{-1})$, $\forall \mathbf{k} \leq \mathbf{n}$ and $r^{-1}(\mathbf{n})k_2^{-1} \rightarrow 0$ as $n \rightarrow \infty$, for every $\mathbf{k} \geq \mathbf{n}$. Since $U_{12}^*(\mathbf{k})$ is a generalized U -statistic, the proof of (4.16) and (4.17) for $\{U_{12}^*(\mathbf{k})\}$ follows directly from (3.15) and the fact that as in Section 3, $E[U_{12}^*(\mathbf{k})]^2 \leq C(\mathbf{F})k_1^{-1}k_2^{-2}$ for every $\mathbf{k} \geq \mathbf{m}$.

Now, we define the sequence of σ -fields $\{\mathcal{E}_{kq}, k \geq m_1, q \geq m_2\}$ as in Section 3 [following (3.7)]. Then it follows from (4.15) that for every $k' \geq k$ and $q' \geq q$,

$$(4.18) \quad E[U_{12}^{**}(k, q) | \mathcal{E}_{k'q'}] = U_{12}^{**}(k', q') \quad \text{a.e.}$$

Consequently, following the same method of approach as in the proof of Lemma 3.2, we obtain from Lemma 3.1 that for every $\mathbf{N} \geq \mathbf{n} \geq \mathbf{m}$,

$$(4.19) \quad E[(\max_{1 \leq j \leq 2} \max_{n_j \leq k_j \leq N_j} |U_{12}^{**}(k_1, k_2)|)^2] \\ \leq 16E[E[U_{12}^{**}(\mathbf{n})]^2 - E[U_{12}^{**}(n_1, N_2)]^2 - E[U_{12}^{**}(N_1, n_2)] + E[U_{12}^{**}(\mathbf{N})]^2] \\ \leq 16E[U_{12}^{**}(\mathbf{n})]^2 \leq 16C(\mathbf{F})n_1^{-1}n_2^{-2}.$$

Thus, the proof of (4.16) and (4.17) for $\{U_{12}^{**}(\mathbf{k})\}$ follows immediately from (4.19) and the Chebychev inequality. \square

5. Weak convergence for random indices. We now consider the case where in (1.3) we allow $(\lambda_1, \dots, \lambda_c)' = \boldsymbol{\lambda}$ to be a stochastic vector with positive elements. More precisely, let $\{\mathbf{N}_n = (N_n^{(1)}, \dots, N_n^{(c)})', n \geq 1\}$ be a sequence of vectors with positive integer valued random variables, such that

$$(5.1) \quad n^{-1}\mathbf{N}_n \rightarrow_p \boldsymbol{\lambda} = (\lambda_1, \dots, \lambda_c)', \quad \text{as } n \rightarrow \infty,$$

where $\lambda_j, j = 1, \dots, c$ are positive random variables defined on the same probability space as of the original $\{X_{ji}, i \geq 1\}, j = 1, \dots, c$.

We define $W(\mathbf{N}_n)$ as in (2.6) when $\mathbf{N}_n \geq \mathbf{1}$; otherwise, we let $W(\mathbf{N}_n) = 0$. Similarly, we define $W^*(\mathbf{N}_n)$ as in (2.9) when $\mathbf{N}_n \geq \mathbf{1}$; otherwise, we let $W^*(\mathbf{N}_n) = 0$. Finally, we define $\tilde{W}(\mathbf{N}_n)$ and $\tilde{W}^*(\mathbf{N}_n)$ as in Section 4, with n being replaced by \mathbf{N}_n . Our basic problem is to study the weak convergence of $W(\mathbf{N}_n), W^*(\mathbf{N}_n), \tilde{W}(\mathbf{N}_n)$ and $\tilde{W}^*(\mathbf{N}_n)$, when $n \rightarrow \infty$ and (5.1) holds.

THEOREM 5.1. *Under (2.3), (2.4) and (5.1), $W(\mathbf{N}_n)$ and $W^*(\mathbf{N}_n)$ converge in law in the extended Skorokhod J_1 -topology on D_c to W and W^* , respectively, defined in (2.8) and (2.12), while under (2.3), (4.3) and (5.1), $\tilde{W}(\mathbf{N}_n)$ and $\tilde{W}^*(\mathbf{N}_n)$ weakly converge to W and W^* respectively.*

PROOF. We shall only consider the case of $W(\mathbf{N}_n)$, as the other cases follow similarly. By Theorem 2.1 and Lemma 3.2, we can verify the conditions of Theorem 2 of Mogyorodi (1967) which insures that the finite dimensional distributions of $\{W(\mathbf{N}_n)\}$ converge to those of W . To establish the tightness of $\{W(\mathbf{N}_n)\}$, we define the *supremum metric* $\rho(x, y) = \sup_{\mathbf{t} \in E_c} |x(\mathbf{t}) - y(\mathbf{t})|$, and the *modulus of continuity* (for $\delta > 0$) $\omega_\delta(x) = \sup\{|x(\mathbf{t}) - x(\mathbf{s})| : \mathbf{t}, \mathbf{s} \in E_c, |\mathbf{t} - \mathbf{s}| < \delta\}$. Now, using our Lemma 3.2, (3.11) and (3.12) in place of Theorem 1 of Wichura (1969) and thereby extending his (2a) to our statistics, we obtain by the same technique as in his proof of Theorem 3 (on pages 686–687) that under (1.3), (2.3) and (2.4), for every $\varepsilon > 0$, as $n \rightarrow \infty$,

$$(5.2) \quad \lim_{\delta \downarrow 0} \limsup_n P\{\omega_\delta(W(\mathbf{n})) > \varepsilon\} = 0.$$

Consider now a sequence of generalized U -statistics

$$(5.3) \quad \begin{aligned} U'([\mathbf{nt}]) &= \prod_{j=1}^c \binom{n_j}{m_j}^{-1} \sum_{(\mathbf{nt})}^{**} g^*(X_{j\alpha_{j_i}}, 1 \leq i \leq m_j, 1 \leq j \leq c) \\ & \quad t_j \geq n_j^{-1}k_n^{(j)}, \quad 1 \leq j \leq c; \\ & = 0, \quad \text{if } t_j < n_j^{-1}k_n^{(j)} \text{ for some } j (= 1, \dots, c), \end{aligned}$$

where $g^*(x_{11}, \dots, x_{cm_c}) = g(x_{11}, \dots, x_{cm_c}) - \theta(\mathbf{F})$, the summation $\sum_{(\mathbf{nt})}^{**}$ extends over all $k_n^{(j)} + 1 \leq \alpha_{j1} < \dots < \alpha_{jm_j} \leq [n_j t_j]$, and $\lim_{n \rightarrow \infty} k_n^{(j)} = \infty$ but $\lim_{n \rightarrow \infty} n^{-1}k_n^{(j)} = 0, 1 \leq j \leq c$. Then, on replacing $U([\mathbf{nt}]) - \theta(\mathbf{F})$ in (2.6) by $U'([\mathbf{nt}]), \mathbf{t} \in E_c$, we define a parallel process $W'(\mathbf{n}) = \{W'(\mathbf{t}; \mathbf{n}) : \mathbf{t} \in E_c\}$.

LEMMA 5.2. *Under (1.3), (2.3) and (2.4), $\rho(W(\mathbf{n}), W'(\mathbf{n})) \rightarrow 0$ almost surely as $n \rightarrow \infty$.*

PROOF. Here also, we let $c = 2$. For $\mathbf{0} \leq \mathbf{l} \leq \mathbf{m} \leq \mathbf{k}_n$, $\mathbf{k} - \mathbf{k}_n \geq \mathbf{m} - \mathbf{l}$, we define

$$(5.4) \quad U_1(\mathbf{k}, \mathbf{k}_n) = \left\{ \prod_{j=1}^2 \binom{k_n^{(j)}}{l_j} \binom{k_j - k_n^{(j)}}{m_j - l_j} \right\}^{-1} \\ \times \sum_{(\mathbf{k})}^* g^*(X_{j\alpha_{ji}}, 1 \leq i \leq m_j, 1 \leq j \leq 2),$$

where the summation $\sum_{(\mathbf{k})}^*$ extends over all $1 \leq \alpha_{j1} < \dots < \alpha_{jl_j} \leq k_n^{(j)} < \alpha_{j(l_j+1)} < \dots \leq \alpha_{jm_j} \leq k_j$, $1 \leq j \leq 2$. Thus $U_1(\mathbf{k}, \mathbf{k}_n)$ may be interpreted as a generalized U -statistic based on four samples of sizes $(k_n^{(1)}, k_1 - k_n^{(1)}, k_n^{(2)}, k_2 - k_n^{(2)})$. Consequently, as in our Lemma 3.2, (3.11), (3.12), (3.13), (3.14) and (3.15), it can be shown that for every $\varepsilon > 0$, there exist a positive $K_\varepsilon (< \infty)$ and an $n_0(\varepsilon)$, such that for $n \geq n_0(\varepsilon)$, under (1.3) and (2.4),

$$(5.5) \quad P(\max_{\mathbf{0} \leq \mathbf{l} \leq \mathbf{m}} \sup_{\mathbf{k} > \mathbf{k}_n} |U_1(\mathbf{k}, \mathbf{k}_n)| > K_\varepsilon) < \varepsilon.$$

Also, by (1.2), (5.3) and (5.4), we obtain that for $[\mathbf{nt}] \geq \mathbf{k}_n$,

$$(5.6) \quad U([\mathbf{nt}]) - \theta(\mathbf{F}) \\ = \sum_{\mathbf{l}=\mathbf{0}}^{\mathbf{m}} \prod_{j=1}^2 \left\{ \binom{m_j}{l_j} \binom{k_j - k_n^{(j)}}{m_j - l_j} \binom{k_n^{(j)}}{l_j} \binom{k_j}{m_j}^{-1} \right\} U_1([\mathbf{nt}], \mathbf{k}_n),$$

where we let $k_j = [n_j t_j]$, $j = 1, 2$. Consequently, by (5.5) and (5.6), as $n \rightarrow \infty$,

$$(5.7) \quad \sup_n \max_{\mathbf{k}_n \leq \mathbf{k} \leq \mathbf{n}} \{n^{-\frac{1}{2}}[\min(k_1, k_2)]|U(\mathbf{k}) - U'(\mathbf{k})|\} \\ \leq \{\sup_n \max_{\mathbf{k}_n \leq \mathbf{k} \leq \mathbf{n}} \max_{\mathbf{0} \leq \mathbf{l} \leq \mathbf{m}} |U_1(\mathbf{k}, \mathbf{k}_n)|\} \{O([n^{-1}(k_n^{(1)})^2], [n^{-1}(k_n^{(2)})^2])\} \\ = O_p((n^{-\frac{1}{2}}k_n^{(1)})^2, (n^{-\frac{1}{2}}k_n^{(2)})^2) = o_p(1).$$

Thus, by (2.6), (2.7) (3.39) and (5.7), as $n \rightarrow \infty$,

$$(5.8) \quad \sup_{n^{-1}\mathbf{k}_n \leq \mathbf{t} \leq \mathbf{1}} |W(\mathbf{t}; \mathbf{n}) - W'(\mathbf{t}; \mathbf{n})| \rightarrow 0, \quad \text{almost surely (a.s.)}$$

On the other hand, for $\mathbf{t} \in E_c^{(n)} = E_c - \{\mathbf{t} : n^{-1}\mathbf{k}_n \leq \mathbf{t} \leq \mathbf{1}\}$, $W'(\mathbf{t}; \mathbf{n}) = 0$ [by (5.3)], so that

$$(5.9) \quad \sup_{\mathbf{t} \in E_c^{(n)}} |W(\mathbf{t}; \mathbf{n}) - W'(\mathbf{t}; \mathbf{n})| \\ = \sup_{\mathbf{t} \in E_c^{(n)}} |W(\mathbf{t}; \mathbf{n})| \\ \leq \{\max_{\mathbf{m} \leq \mathbf{k} \leq \mathbf{n}} |U(\mathbf{k}) - \theta(\mathbf{F})|\} \{\sup_{\mathbf{t} \in E_c^{(n)}} \phi(\{\mathbf{nt}\}, n)\} \\ \leq \{\sup_{\mathbf{k} \leq \mathbf{m}} |U(\mathbf{k}) - \theta(\mathbf{F})|\} \{O([n^{-\frac{1}{2}}k_n^{(1)}]^2, [n^{-\frac{1}{2}}k_n^{(2)}]^2)\} \\ = O(1) \cdot o(1) = o(1), \quad \text{a.s.,} \quad \text{as } n \rightarrow \infty.$$

The lemma follows from (5.8) and (5.9).

LEMMA 5.3. *If $A \in \mathcal{A}$, then for every $\varepsilon > 0$ and $\eta > 0$, there exists a $\delta (> 0)$, sufficiently small, such that $\limsup_n P\{\omega_\delta(W'(\mathbf{n})) > \varepsilon | A\} < \eta$.*

PROOF. By (5.2) and Lemma 5.2, for every $\varepsilon > 0$,

$$\lim_{\delta \downarrow 0} \limsup_n P\{\omega_\delta(W'(\mathbf{n})) > \varepsilon\} < \eta.$$

Also, by definition in (5.3), $W'(\mathbf{n})$ depends only on the set $\{X_{ji}, k_n^{(j)} < i < n_j, j = 1, \dots, c\}$. Hence, using Rényi's (1958) idea of mixing sequence of sets and

proceeding as in Lemma 3 of Blum, Hanson and Rosenblatt (1963), the result follows. \square

We return now to the proof of Theorem 5.1. By virtue of (5.1), for every $\delta' > 0$, $P\{|n^{-1}\mathbf{N}_n - \boldsymbol{\lambda}| > \delta'\} \rightarrow 0$ as $n \rightarrow \infty$. As such, it can be shown by standard steps [as in Mogyorodi (1967)] that for every $\varepsilon > 0$, there exist an $\eta > 0$ and an integer n_0 , such that $P\{\rho(W(\mathbf{N}_n), W'(\mathbf{N}_n)) > \varepsilon\} < \eta$ for $n \geq n_0$. Hence, to establish the tightness of $\{W(\mathbf{N}_n)\}$, it suffices to show that for every $\varepsilon > 0$,

$$(5.10) \quad \limsup_n P\{\omega_\delta(W'(\mathbf{N}_n)) > \varepsilon\} \rightarrow 0 \quad \text{as } \delta \downarrow 0.$$

We note that by (1.2), (2.6), (2.7), (5.2), (5.3), (5.6) and Lemma 5.2, for every $\varepsilon > 0$ and $\eta > 0$, there exist a $\delta > 0$ and an $n_0(\varepsilon, \eta)$, such that for $n \geq n_0(\varepsilon, \eta)$,

$$(5.11) \quad P\{\max_{k:|\mathbf{k}-\mathbf{n}|<\delta n} \rho(W'(\mathbf{k}), W'(\mathbf{n})) > \varepsilon\} < \eta.$$

Hence, on using the inequality that $\omega_\delta(W'(\mathbf{N}_n)) \leq 2\rho(W'(\mathbf{N}_n), W'([n\boldsymbol{\lambda}])) + \omega_\delta(W([n\boldsymbol{\lambda}]))$, it suffices to show that

$$(5.12) \quad \limsup_n P\{\omega_\delta(W'([n\boldsymbol{\lambda}])) > \varepsilon\} \rightarrow 0 \quad \text{as } \delta \downarrow 0,$$

$$(5.13) \quad P\{\rho(W'(\mathbf{N}_n), W'([n\boldsymbol{\lambda}])) > \varepsilon\} \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Consider the events $A(\mathbf{h}) = \{\boldsymbol{\lambda}: a_0(\eta) + h_j \delta' < \lambda_j \leq a_0(\eta) + (h_j + 1)\delta', 1 \leq j \leq c\}$, $\mathbf{h} \geq \mathbf{0}$, $a_0(\eta) > 0$, and let $\mathbf{a}(\mathbf{h}) = (a_0(\eta) + (h_j + \frac{1}{2})\delta', 1 \leq j \leq c)$, $\mathbf{h} \geq \mathbf{0}$. Then, we have

$$(5.14) \quad P\{\omega_\delta(W'([n\boldsymbol{\lambda}])) > \varepsilon\} \leq P\{\min_{1 \leq j \leq c} \lambda_j \leq a_0(\eta)\} + \sum_{\mathbf{h}=\mathbf{0}}^\infty P\{A(\mathbf{h})\}P\{\omega_\delta(W'([n\boldsymbol{\lambda}])) > \varepsilon | A(\mathbf{h})\},$$

where we let $P(B|A) = 0$ when $P(A) = 0$. Since, when $A(\mathbf{h})$ holds, $|\boldsymbol{\lambda} - \mathbf{a}(\mathbf{h})| < \delta'$, it can be shown on using (5.11) and Lemma 5.3 that for every $\varepsilon > 0$ and $\eta > 0$, there exists a $\delta'(0 < \delta' < \frac{1}{2}a_0(\eta))$, and an $n_0(\varepsilon, \eta)$, such that for $n \geq n_0(\varepsilon, \eta)$ and all $\mathbf{h} \geq \mathbf{0}$, $P\{\omega_\delta(W'([n\boldsymbol{\lambda}])) > \varepsilon | A(\mathbf{h})\} < \eta/2, \forall \mathbf{h} > \mathbf{0}$. Hence (5.12) holds. Finally, on writing

$$(5.15) \quad \begin{aligned} &P\{\rho(W'(\mathbf{N}_n), W'([n\boldsymbol{\lambda}])) > \varepsilon\} \\ &\leq P\{|n^{-1}\mathbf{N}_n - \boldsymbol{\lambda}| > \delta'\} + P\{\min_{1 \leq j \leq c} \lambda_j \leq a_0(\eta)\} + \sum_{\mathbf{h}=\mathbf{0}}^\infty P\{A(\mathbf{h})\} \\ &\quad P\{\rho(W'(\mathbf{N}_n), W'([n\boldsymbol{\lambda}])) > \varepsilon, |n^{-1}\mathbf{N}_n - \boldsymbol{\lambda}| \leq \delta' | A(\mathbf{h})\}, \end{aligned}$$

the proof of (5.13) follows on parallel lines.

The theorem is useful in the context of sequential procedures based on generalized U -statistics; we may refer to Williams and Sen (1973) for such an application.

As an illustration of the uses of Theorems 2.1, 2.2, 4.1 and 5.1, we consider a simple case where $p = 1, c = 2$ and

$$(5.16) \quad \theta(F_1, F_2) = \int_{-\infty}^\infty F_1(x) dF_2(x) = P\{X_{1i} \leq X_{2j}\},$$

and we assume that both F_1 and F_2 are continuous everywhere. Then

$\theta(F(\cdot, \mathbf{n})) = U(\mathbf{n})$ is the Wilcoxon rank statistic

$$(5.17) \quad \frac{1}{n_1 n_2} \sum_{i=1}^{n_1} \sum_{j=1}^{n_2} c(X_i, X_j); \quad c(u, v) = 1, \quad u \leq v \\ = 0, \quad u > v.$$

Here $m_1 = m_2 = 1$, and the summands $\{c(X_i, X_j), 1 \leq i \leq n, 1 \leq j \leq n_2\}$ are not all independent, so that Wichura's (1969) results do not hold. If the two distributions F_1 and F_2 are mutually overlapping, then (2.3) holds, while (2.4) holds for all F_1, F_2 , as $c(u, v)$ is bounded. If $F_1 \equiv F_2$, $\theta(F_1, F_2) = \frac{1}{2}$, $\sigma_1^2 = \sigma_2^2 = \frac{1}{12}$, so that the results further simplify. Theorem 5.1 for the Wilcoxon statistic is useful for the problem of sequential testing and estimating $\theta(F_1, F_2)$.

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