

SOME EXTENSIONS OF THE WALD-WOLFOWITZ-NOETHER THEOREM

BY JAROSLAV HÁJEK

Mathematical Institute of the Czechoslovak Academy of Sciences

1. Summary and introduction. Let $(R_{\nu 1}, \dots, R_{\nu N_\nu})$ be a random vector which takes on the $N_\nu!$ permutations of $(1, \dots, N_\nu)$ with equal probabilities. Let $\{b_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ and $\{a_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ be double sequences of real numbers. Put

$$(1.1) \quad S_\nu = \sum_{i=1}^{N_\nu} b_{\nu i} a_{\nu R_{\nu i}}.$$

We shall prove that the sufficient and necessary condition for asymptotic $(N_\nu \rightarrow \infty)$ normality of S_ν is of Lindeberg type. This result generalizes previous results by Wald-Wolfowitz [1], Noether [3], Hoeffding [4], Dwass [6], [7] and Motoo [8]. In respect to Motoo [8] we show, in fact, that his condition, applied to our case, is not only sufficient but also necessary.

Cases encountered in rank-test theory are studied in more detail in Section 6 by means of the theory of martingales. The method of this paper consists in proving asymptotic equivalency in the mean of (1.1) to a sum of infinitesimal independent components.

2. Three lemmas. Consider a sequence U_1, \dots, U_N of independent random variables each having uniform (rectangular) distribution over the interval $(0, 1]$. Let R_i be the rank of U_i , i.e.,

$$(2.1) \quad U_i = Z_{R_i},$$

where $Z_1 < \dots < Z_N$ is the sequence U_1, \dots, U_N , reordered in ascending magnitude.

Take a nondecreasing sequence $a_1 \leq \dots \leq a_N$ of real numbers and put

$$(2.2) \quad a(\lambda) = a_i \quad \text{for} \quad (i-1)/N < \lambda \leq i/N \quad (1 \leq i \leq N).$$

The function $a(\lambda)$ will be called a quantile function of $a_1 \leq \dots \leq a_N$. As $(i-1)/N < i/(N+1) < i/N$, we have

$$(2.3) \quad a_i = a(i/N) = a[i/(N+1)].$$

Furthermore,

$$(2.4) \quad \begin{aligned} \bar{a} = \frac{1}{N} \sum_{i=1}^N a_i &= \int_0^1 a(\lambda) d\lambda \quad \text{and} \quad \sigma^2 = \frac{1}{N} \sum_{i=1}^N (a_i - \bar{a})^2 \\ &= \int_0^1 [a(\lambda) - \bar{a}]^2 d\lambda. \end{aligned}$$

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LEMMA 2.1.

$$(2.5) \quad E \left[a(U_1) - a \left(\frac{R_1}{N} \right) \right]^2 \leq 2 \max_{1 \leq i \leq N} |a_i - \bar{a}| \frac{1}{N} \left[2 \sum_{i=1}^N (a_i - \bar{a})^2 \right]^{\frac{1}{2}},$$

where the function $a(\cdot)$ is given by (2.2).

PROOF. If Z_1, \dots, Z_N are fixed, then U_1 takes on each of the values Z_i with probability $1/N$, and $U_1 = Z_i$ is equivalent to $R_1 = i$. Therefore,

$$(2.6) \quad \begin{aligned} E[a(U_1) - a(R_1/N)]^2 &= EE\{[a(U_1) - a(R_1/N)]^2 \mid Z_1, \dots, Z_N\} \\ &= (1/N) E \sum_{i=1}^N [a(Z_i) - a(i/N)]^2. \end{aligned}$$

Now, first, consider a special quantile function

$$(2.7) \quad \begin{aligned} \epsilon[\lambda - (k/N)] &= 0 \quad \text{if } \lambda \leq k/N \\ &= 1 \quad \text{if } \lambda > k/N. \end{aligned}$$

The quantile function $\epsilon[\lambda - (k/N)]$ corresponds, obviously, to the sequence $a_1 = \dots = a_k = 0$, $a_{k+1} = \dots = a_N = 1$. Let K denote the number of the U_i 's smaller than k/N . Clearly, $Z_K < k/N < Z_{K+1}$. If $K \leq k$, we have

$$(2.8) \quad \begin{aligned} \epsilon[Z_i - (k/N)] - \epsilon[(i - k)/N] &= 0 \quad \text{if } i = 1, \dots, K, k+1, \dots, N \\ &= 1 \quad \text{otherwise} \end{aligned}$$

so that

$$(2.9) \quad \sum_{i=1}^N \left[\epsilon \left(Z_i - \frac{k}{N} \right) - \epsilon \left(\frac{i - k}{N} \right) \right]^2 = |K - k|.$$

We can easily see that (2.9) also holds for $K \geq k$. The result (2.9) together with (2.6) gives

$$(2.10) \quad E \left[\epsilon \left(U_1 - \frac{k}{N} \right) - \epsilon \left(\frac{R_1 - k}{N} \right) \right]^2 = \frac{1}{N} E |K - k|.$$

The distribution of K is, obviously, binomial with mean value k and variance $k[1 - (k/N)]$, so that

$$E|K - k| \leq [E(K - k)^2]^{\frac{1}{2}} = [k\{1 - (k/N)\}]^{\frac{1}{2}},$$

and, therefore,

$$(2.11) \quad E\{\epsilon(U_1 - k/N) - \epsilon[(R_1 - k)/N]\}^2 \leq N^{-1}\{k[1 - (k/N)]\}^{\frac{1}{2}}.$$

Now let us only suppose that $a_1 = 0$, and otherwise the sequence $a_1 \leq \dots \leq a_N$ can be arbitrary. The quantile function of any such sequence may be expressed in the form

$$(2.12) \quad a(\lambda) = \sum_{k=1}^{N-1} (a_{k+1} - a_k) \epsilon[\lambda - (k/N)] \quad (a_1 = 0, 0 < \lambda \leq 1).$$

Actually, e.g., for $\lambda = i/N$, we have

$$\alpha(i/N) = \sum_{k=1}^{N-1} (a_{k+1} - a_k) \epsilon[(i-k)/N] = \sum_{k=1}^{i-1} (a_{k+1} - a_k) = a_i$$

$$(1 \leq i \leq N).$$

Now from (2.12) it follows that, first,

$$\begin{aligned} \sum_{i=1}^N a_i^2 &= \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j) \sum_{i=1}^N \epsilon[(i-k)/N] \epsilon[(i-j)/N] \\ (2.13) \quad &= \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j)(N - \max(k, j)), \end{aligned}$$

and, second,

$$\begin{aligned} \left[a(Z_i) - a\left(\frac{i}{N}\right) \right]^2 &= \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j) \\ (2.14) \quad &\cdot \left[\epsilon\left(Z_i - \frac{k}{N}\right) - \epsilon\left(\frac{i-k}{N}\right) \right] \left[\epsilon\left(Z_i - \frac{j}{N}\right) - \epsilon\left(\frac{i-j}{N}\right) \right]. \end{aligned}$$

Because $\epsilon[Z_i - (k/N)] = \epsilon[(i-k)/N]$ and $\epsilon[Z_i - (j/N)] = \epsilon[(i-j)/N]$ take on only the values 0 and ± 1 , we have

$$\begin{aligned} \left[\epsilon\left(Z_i - \frac{k}{N}\right) - \epsilon\left(\frac{i-k}{N}\right) \right] \left[\epsilon\left(Z_i - \frac{j}{N}\right) - \epsilon\left(\frac{i-j}{N}\right) \right] \\ (2.15) \quad \leq \left[\epsilon\left(Z_i - \frac{\max(k, j)}{N}\right) - \epsilon\left(\frac{i - \max(k, j)}{N}\right) \right]^2. \end{aligned}$$

On combining (2.14) (2.15) and (2.11), we get

$$\begin{aligned} E \left[a(U_1) - a\left(\frac{R_1}{N}\right) \right]^2 &= \frac{1}{N} E \sum_{i=1}^N \left[a(Z_i) - a\left(\frac{i}{N}\right) \right]^2 \\ &\leq \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j) \\ &\quad \cdot \frac{1}{N} E \sum_{i=1}^N \left[\epsilon\left(Z_i - \frac{\max(k, j)}{N}\right) - \epsilon\left(\frac{i - \max(k, j)}{N}\right) \right]^2 \\ &= \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j) E \left[\epsilon\left(U_1 - \frac{\max(k, j)}{N}\right) - \epsilon\left(\frac{R_1 - \max(k, j)}{N}\right) \right]^2 \\ &\leq \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j) \frac{1}{N} \left[\max(k, j) \frac{N - \max(k, j)}{N} \right]^{\frac{1}{2}} \\ &\leq \frac{1}{N} \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j) [N - \max(k, j)]^{\frac{1}{2}} \end{aligned}$$

$$\leq \frac{1}{N} \left\{ \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j) \cdot \sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j)(N - \max(k, j)) \right\}^{\frac{1}{2}}.$$

According to (2.13) and to

$$\sum_{k=1}^{N-1} \sum_{j=1}^{N-1} (a_{k+1} - a_k)(a_{j+1} - a_j) = a_N^2$$

the last expression equals $N^{-1} a_N (\sum_{i=1}^N a_i^2)^{\frac{1}{2}}$. This means that, for $a_1 = 0$,

$$(2.16) \quad E \left[a(U_1) - a \left(\frac{R_1}{N} \right) \right]^2 \leq \frac{1}{N} a_N \left[\sum_{i=1}^N a_i^2 \right]^{\frac{1}{2}}.$$

Generally, for an arbitrary a_1 , we have

$$(2.17) \quad E \left[a(U_1) - a \left(\frac{R_1}{N} \right) \right]^2 \leq \frac{1}{N} (a_N - a_1) \left[\sum_{i=1}^N (a_i - a_1)^2 \right]^{\frac{1}{2}}.$$

On making use of the values $-a_N \leq \dots \leq -a_1$ instead of $a_1 \leq \dots \leq a_N$, we also get from (2.17) that

$$(2.18) \quad E \left[a(U_1) - a \left(\frac{R_1}{N} \right) \right]^2 \leq \frac{1}{N} (a_N - a_1) \left[\sum_{i=1}^N (a_i - a_N)^2 \right]^{\frac{1}{2}}.$$

It is now easy to derive (2.4). Let us put

$$\begin{aligned} a^+(\lambda) &= \bar{a} & \text{if } a(\lambda) \leq \bar{a} \\ &= a(\lambda) & \text{if } a(\lambda) \geq \bar{a} \end{aligned}$$

and

$$\begin{aligned} a^-(\lambda) &= a(\lambda) - \bar{a} & \text{if } a(\lambda) \leq \bar{a} \\ &= 0 & \text{if } a(\lambda) \geq \bar{a}. \end{aligned}$$

We then have

$$a(\lambda) = a^+(\lambda) + a^-(\lambda).$$

and, in view of (2.17) and (2.18), the following inequalities are clear:

$$\begin{aligned} E \left[a(U_1) - a \left(\frac{R_1}{N} \right) \right]^2 &\leq 2E \left[a^+(U_1) - a^+ \left(\frac{R_1}{N} \right) \right]^2 + 2E \left[a^-(U_1) - a^- \left(\frac{R_1}{N} \right) \right]^2 \\ &\leq 2(a_N - \bar{a}) \left[\sum_{a_i \geq \bar{a}} (a_i - \bar{a})^2 \right]^{\frac{1}{2}} + 2(\bar{a} - a_1) \left[\sum_{a_i \leq \bar{a}} (a_i - \bar{a})^2 \right]^{\frac{1}{2}} \\ &\leq 2 \max_{1 \leq i \leq N} |a_i - \bar{a}| \left\{ \left[\sum_{a_i \geq \bar{a}} (a_i - \bar{a})^2 \right]^{\frac{1}{2}} + \left[\sum_{a_i \leq \bar{a}} (a_i - \bar{a})^2 \right]^{\frac{1}{2}} \right\} \\ &\leq 2 \max_{1 \leq i \leq N} |a_i - \bar{a}| \left[2 \sum_{i=1}^N (a_i - \bar{a})^2 \right]^{\frac{1}{2}}. \end{aligned}$$

This completes the proof.

Let us have a nondecreasing quadratically integrable function $\varphi(\lambda)$, $0 < \lambda < 1$, and put

$$(2.19) \quad \varphi_N(\lambda) = \varphi[i/(N+1)] \quad \text{if} \quad (i-1)/N < \lambda \leq i/N.$$

LEMMA 2.2. *The functions $\varphi_N^2(\lambda)$, $0 < \lambda < 1$ are uniformly ($N \geq 1$) integrable and*

$$(2.20) \quad \lim_{N \rightarrow \infty} \int_0^1 [\varphi_N(\lambda) - \varphi(\lambda)]^2 d\lambda = 0.$$

PROOF. Suppose that $\varphi(0) \geq 0$ so that both $\varphi(\lambda)$ and $\varphi^2(\lambda)$ are non-decreasing. Let A be a subset of $[0, 1]$ and $\mu(A)$ its Lebesgue measure. Put

$$I_k = ((k-1)/N, 1/N)$$

and $J_k = (k/N - \mu(A \cap I_k), k/N)$ and note that $\varphi_N(\lambda) = \varphi(k/(N+1))$ for $\lambda \in I_k$ and $\varphi_N(\lambda) \leq \varphi(\lambda)$ for $k/(N+1) \leq \lambda < k/N$. It holds, obviously,

$$(2.21) \quad \int_{A \cap I_k} \varphi_N^2(\lambda) d\lambda = \mu(A \cap I_k) \varphi^2(k/(N+1)) \leq (N+1)k^{-1} \int_{J_k} \varphi^2(\lambda) d\lambda$$

so that, on making use of the first right hand expression for $k \leq 3N/4$ and of the second one for $k \geq N/4$, we get

$$(2.22) \quad \int_A \varphi_N^2(\lambda) d\lambda = \sum_{k=1}^N \int_{A \cap I_k} \varphi_N^2(\lambda) d\lambda \leq \varphi^2(3/4) \mu(A) + 4 \int_B \varphi^2(\lambda) d\lambda$$

where $\mu(B) = \mu(A)$. Inequality (2.22) clearly proves uniform integrability of the functions $\{\varphi_N^2(\lambda)\}$. A general $\varphi(\lambda)$ may be written in the form $\varphi(\lambda) = \varphi_1(\lambda) - \varphi_2(1-\lambda)$, where $\varphi_1(0) \geq 0$, $\varphi_2(0) \geq 0$ and both $\varphi_1(\lambda)$ and $\varphi_2(\lambda)$ are non-decreasing. This completes the proof of uniform integrability.

In order to prove (2.20), let us observe that $\varphi_N(\lambda) \rightarrow \varphi(\lambda)$ on the set of continuity points of $\varphi(\lambda)$, which, however, has Lebesgue measure 1. Convergence of $\varphi_N(\lambda)$ to $\varphi(\lambda)$ almost everywhere, together with uniform integrability of the function $\varphi_N^2(\lambda)$, implies (2.20). The proof is completed.

LEMMA 2.3. *Let $c_1, \dots, c_N, d_1, \dots, d_N$ be arbitrary real numbers and put $\bar{c} = N^{-1} \sum_{i=1}^N c_i$, $\bar{d} = N^{-1} \sum_{i=1}^N d_i$. Then*

$$(2.23) \quad \begin{aligned} \text{var} \left(\sum_{i=1}^N c_i d_{R_i} \right) &= \frac{1}{N-1} \sum_{i=1}^N (c_i - \bar{c})^2 \sum_{i=1}^N (d_i - \bar{d})^2 \\ &\leq \frac{1}{N-1} \sum_{i=1}^N (c_i - \bar{c})^2 \sum_{i=1}^N d_i^2. \end{aligned}$$

The proof is immediate.

3. Asymptotic equivalency in the mean. Random variables S_ν and T_ν will be called asymptotically equivalent in the mean (symbolically, $S_\nu \sim T_\nu$)

$$(3.1) \quad \lim_{\nu \rightarrow \infty} E(S_\nu - T_\nu)^2 / (\text{var } T_\nu) = 0.$$

The relation is symmetric, and, if $S_\nu \sim T_\nu$ and $T_\nu \sim V_\nu$, then, clearly, $S_\nu \sim V_\nu$.

Let us take a sequence of independent random variables U_1, U_2, \dots each uniformly distributed over $(0, 1]$ and denote the rank of U_i in the partial sequence U_1, \dots, U_{N_ν} by $R_{\nu i}$, $1 \leq i \leq N_\nu$, $\nu \geq 1$. The partial sequence U_1, \dots, U_{N_ν} , reordered in ascending magnitude, will be denoted by $Z_{\nu 1}, \dots, Z_{\nu N_\nu}$.

The distribution of S_ν given by (1.1) does not depend on the ordering of the a 's. So we may suppose that

$$(3.2) \quad a_{\nu 1} \leq \dots \leq a_{\nu N_\nu} \quad (\nu \geq 1).$$

We shall assume that the a 's fulfill the condition

$$(3.3) \quad \lim_{\nu \rightarrow \infty} \frac{\max_{1 \leq i \leq N_\nu} (a_{\nu i} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} = 0.$$

THEOREM 3.1. *Under the assumptions (3.2) and (3.3) the statistic S_ν given by (1.1) is asymptotically equivalent in the mean to the statistic*

$$(3.4) \quad T_\nu = \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu) a_\nu(U_i) + \bar{b}_\nu \sum_{i=1}^{N_\nu} a_{\nu i},$$

where $a_\nu(\cdot)$ denotes the quantile functions of $a_{\nu 1} \leq \dots \leq a_{\nu N_\nu}$ given by (2.2), and $\bar{b}_\nu = N_\nu^{-1} \sum_{i=1}^{N_\nu} b_{\nu i}$.

PROOF. On making use of (2.1) and (2.3), we may write

$$(3.5) \quad S_\nu - T_\nu = \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu) \left[a_\nu(Z_{\nu R_{\nu i}}) - a_\nu\left(\frac{R_{\nu i}}{N_\nu}\right) \right].$$

As is well-known, the distribution of the ranks $(R_{\nu 1}, \dots, R_{\nu N_\nu})$ is independent of the vector $(Z_{\nu 1}, \dots, Z_{\nu N_\nu})$. In view of Lemma 2.3, where we put

$$c_i = b_{\nu i} - \bar{b}_\nu$$

and $d_i = a_\nu(Z_{\nu i}) - a_\nu(i/N)$, we can write

$$(3.6) \quad \begin{aligned} E\{(S_\nu - T_\nu)^2 | Z_{\nu 1}, \dots, Z_{\nu N_\nu}\} &= \text{var} \{S_\nu - T_\nu | Z_{\nu 1}, \dots, Z_{\nu N_\nu}\} \\ &\leq \frac{1}{N_\nu - 1} \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{i=1}^{N_\nu} \left[a_\nu(Z_{\nu R_{\nu i}}) - a_\nu\left(\frac{R_{\nu i}}{N_\nu}\right) \right]^2 \\ &= \frac{1}{N_\nu - 1} \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{i=1}^{N_\nu} \left[a_\nu(U_i) - a_\nu\left(\frac{R_{\nu i}}{N_\nu}\right) \right]^2. \end{aligned}$$

The first equality in (3.6) is ensured by $E(S_\nu - T_\nu | Z_{\nu 1}, \dots, Z_{\nu N_\nu}) = 0$ which follows from $\sum (b_{\nu i} - \bar{b}_\nu) = 0$. Taking the mean value over $Z_{\nu 1}, \dots, Z_{\nu N_\nu}$, on both sides of the inequality (3.6), we obtain

$$(3.7) \quad E(S_\nu - T_\nu)^2 \leq \frac{1}{N_\nu - 1} \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{i=1}^{N_\nu} E \left[a_\nu(U_i) - a_\nu\left(\frac{R_{\nu i}}{N_\nu}\right) \right]^2.$$

Clearly,

$$(3.8) \quad E \left[a_\nu(U_i) - a_\nu\left(\frac{R_{\nu i}}{N_\nu}\right) \right]^2 = E \left[a_\nu(U_1) - a_\nu\left(\frac{R_{\nu 1}}{N_\nu}\right) \right]^2 \quad (1 \leq i \leq N_\nu, \nu \geq 1),$$

so that (3.7) may be put in the form

$$(3.9) \quad E(S_\nu - T_\nu)^2 \leq \frac{N_\nu}{N_\nu - 1} E \left[a_\nu(U_1) - a_\nu \left(\frac{R_{\nu 1}}{N_\nu} \right) \right]^2 \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2.$$

On the other hand, it follows from (3.4) that

$$(3.10) \quad \text{var } T_\nu = 1/N_\nu \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2.$$

Making use of Lemma 2.1, (3.9) and (3.10) yield

$$(3.11) \quad \frac{E(S_\nu - T_\nu)^2}{\text{var } T_\nu} \leq \frac{2\sqrt{2} N_\nu}{N_\nu - 1} \frac{\max |a_{\nu i} - \bar{a}_\nu|}{\left[\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \right]^{1/2}}.$$

Consequently, the relation (3.1) follows from (3.11) and (3.3). The proof is completed.

THEOREM 3.2. *Let the function $\varphi(\lambda)$ be non-decreasing, non-constant and quadratically integrable. Suppose that*

$$(3.12) \quad \lim_{\nu \rightarrow \infty} N_\nu = \infty.$$

Then the statistic

$$(3.13) \quad S_\nu = \sum_{i=1}^{N_\nu} b_{\nu i} \varphi[R_{\nu i}/(N_\nu + 1)]$$

and

$$(3.14) \quad T_\nu = \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu) \varphi(U_i) + \bar{b}_\nu \sum_{i=1}^{N_\nu} \varphi[i/(N_\nu + 1)]$$

are asymptotically equivalent in the mean.

PROOF. If we put

$$(3.15) \quad a_{\nu i} = \varphi[i/(N_\nu + 1)],$$

the quantile function of $a_{\nu 1} \leq \dots \leq a_{\nu N_\nu}$ will equal $\varphi_{N_\nu}(\lambda)$ expressed by (2.19). According to Lemma 2.2, the functions $\varphi_{N_\nu}^2(\lambda)$ are uniformly integrable and hence

$$(3.16) \quad \lim_{\nu \rightarrow \infty} N_\nu^{-1} \max_{1 \leq i \leq N_\nu} |a_{\nu i}|^2 = \lim_{\nu \rightarrow \infty} \max_{1 \leq i \leq N_\nu} \int_{(i-1)/N_\nu}^{i/N_\nu} \varphi_{N_\nu}^2(\lambda) d\lambda = 0.$$

On the other hand, from (2.20) and from non-constancy of $\varphi(\lambda)$, it follows that

$$(3.17) \quad \lim_{\nu \rightarrow \infty} \frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 = \int_0^1 \left[\varphi(\lambda) - \int_0^1 \varphi(x) dx \right]^2 d\lambda > 0.$$

Relations (3.16) and (3.17) imply that the sequences $a_{\nu 1} \leq \dots \leq a_{\nu N_\nu}$ fulfill

condition (3.3). Therefore, in view of Theorem 3.1, the statistic (3.13) is asymptotically equivalent to the statistic

$$(3.18) \quad T'_\nu = \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu) \varphi_{N_\nu}(U_i) + \bar{b}_\nu \sum_{i=1}^{N_\nu} \varphi[i/(N_\nu + 1)].$$

It remains to show that (3.18) is equivalent to (3.14). However, as is easy to see,

$$(3.19) \quad \frac{E(T_\nu - T'_\nu)^2}{\text{var } T_\nu} = \frac{\int_0^1 [\varphi(\lambda) - \varphi_{N_\nu}(\lambda)]^2 d\lambda}{\int_0^1 \left[\varphi(\lambda) - \int_0^1 \varphi(x) dx \right]^2 d\lambda},$$

so that $T_\nu \sim T'_\nu$ is a consequence of the assumption (3.12) and of Lemma 2.2. Theorem 3.2 is proved.

4. Necessary and sufficient condition for asymptotic normality of S_ν . If $S_\nu \sim T_\nu$, then obviously, the asymptotic variance and the asymptotic distribution of S_ν and T_ν exist under the same conditions, and, if they exist, are the same. Thus the problem of the asymptotical distribution of S_ν is reduced to the problem of the asymptotical distribution of T_ν . The statistic T_ν , however, is a sum of independent addends, so that, if these addends are infinitesimal, it suffices to use well-known theory [11].

THEOREM 4.1. *Let us suppose that*

$$(4.1) \quad \lim_{\nu \rightarrow \infty} \frac{\max_{1 \leq i \leq N_\nu} (a_{\nu i} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} = 0$$

and

$$(4.2) \quad \lim_{\nu \rightarrow \infty} \frac{\max_{1 \leq i \leq N_\nu} (b_{\nu i} - \bar{b}_\nu)^2}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2} = 0.$$

Then the statistic (1.1) has an asymptotically normal distribution with mean value ES_ν and variance $\text{var } S_\nu$ if, and only if, for any $\tau > 0$

$$(4.3) \quad \lim_{\nu \rightarrow \infty} 1/N_\nu \sum_{i,j} \delta_{\nu ij}^2 = 0,$$

where

$$(4.4) \quad \delta_{\nu ij} = \frac{(b_{\nu i} - \bar{b}_\nu)(a_{\nu j} - \bar{a}_\nu)}{\left[\frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \right]^{1/2}} \quad (1 \leq i, j \leq N_\nu, \nu \geq 1).$$

PROOF. Assuming that $a_{\nu 1} \leq \dots \leq a_{\nu N_\nu}$, then from (4.1) and Theorem 3.1 it follows that (1.1) is asymptotically equivalent in the mean to (3.4). Therefore, it suffices to show that, under the additional assumption (4.2), the condition (4.3) is necessary and sufficient for the asymptotic normality of T_ν with

mean value ET , and variance $\text{var } T$. The assumption (4.2), however, implies that the addends in (3.4) are infinitesimal, because

$$(4.5) \quad \frac{\max_{1 \leq i \leq N_\nu} \text{var} [(b_{\nu i} - \bar{b}_\nu) a_\nu(U_i)]}{\text{var } T_\nu} = \frac{\max_{1 \leq i \leq N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} = \frac{\max_{1 \leq i \leq N_\nu} (b_{\nu i} - \bar{b}_\nu)^2}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2}.$$

Consequently, we have to prove that (4.3) coincides with the Lindeberg condition for T_ν , namely with

$$(4.6) \quad \lim_{\nu \rightarrow \infty} \sum_{i=1}^{N_\nu} \frac{1}{\text{var } T_\nu} \int_{|x| > \tau(\text{var } T_\nu)^{\frac{1}{2}}} x^2 dP \{(b_{\nu i} - \bar{b}_\nu) a_\nu(U_i) < x\} = 0.$$

Clearly, we have

$$(4.7) \quad \frac{1}{\text{var } T_\nu} \int_{|x| > (\text{var } T_\nu)^{\frac{1}{2}}} x^2 dP \{(b_{\nu i} - \bar{b}_\nu) a_\nu(U_i) < x\} = \frac{\frac{1}{N_\nu} \sum_{i=1}^{N_\nu} [(b_{\nu i} - \bar{b}_\nu)^2 \sum_{j \in E_{\nu i}} (a_{\nu j} - \bar{a}_\nu)^2]}{\frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2},$$

where

$$(4.8) \quad E_{\nu i} = \left\{ j : |(b_{\nu i} - \bar{b}_\nu)(a_{\nu j} - \bar{a}_\nu)| > \tau \left[\frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \right]^{\frac{1}{2}} \right\}.$$

Now, observe that

$$(4.9) \quad \frac{1}{N_\nu} \sum_{i=1}^{N_\nu} \sum_{|b_{\nu i} - \bar{b}_\nu| > \tau} \delta_{\nu i j}^2 = \frac{\sum_{i=1}^{N_\nu} [(b_{\nu i} - \bar{b}_\nu)^2 \sum_{j \in E_{\nu i}} (a_{\nu j} - \bar{a}_\nu)^2]}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2},$$

so that (4.3) is actually equivalent to (4.6). The proof is accomplished.

The condition (4.3) is symmetrical in the a 's and the b 's. In applications, however, the a 's and the b 's often play a somewhat different role. In such cases the following theorem is useful:

THEOREM 4.2. *A double sequence $\{a_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ satisfying the condition (4.1) fulfills the condition (4.3) for any double sequence $\{b_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ satisfying the condition (4.2) if, and only if,*

$$(4.10) \quad \left[\lim_{\nu \rightarrow \infty} \frac{k_\nu}{N_\nu} = 0 \right] \Rightarrow \lim_{\nu \rightarrow \infty} \frac{\max_{1 \leq i_1 < \dots < i_{k_\nu} \leq N_\nu} \sum_{\alpha=1}^{k_\nu} (a_{\nu i_\alpha} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} = 0.$$

PROOF. First, let us prove that (4.2) and (4.10) implies (4.3). Put, for a fixed $\tau > 0$,

$$(4.11) \quad E_\nu = \left\{ j: (a_{\nu j} - \bar{a}_\nu)^2 > \frac{\tau^2 \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2}{\max_{1 \leq i \leq N_\nu} (b_{\nu i} - \bar{b}_\nu)^2} \frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \right\}$$

and denote the number of elements in E_ν by k_ν . Clearly

$$\frac{\tau^2 \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2}{\max_{1 \leq i \leq N_\nu} (b_{\nu i} - \bar{b}_\nu)^2} \frac{k_\nu}{N_\nu} \sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2 < \sum_{j \in E_\nu} (a_{\nu j} - \bar{a}_\nu)^2 \leq \sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2,$$

i.e.,

$$\frac{k_\nu}{N_\nu} \leq \tau^{-2} \frac{\max_{1 \leq i \leq N_\nu} (b_{\nu i} - \bar{b}_\nu)^2}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2},$$

from which, in view of (4.2), it follows that

$$(4.12) \quad \lim_{\nu \rightarrow \infty} (k_\nu / N_\nu) = 0.$$

Relation (4.12), according to (4.10), implies that

$$(4.13) \quad \lim_{\nu \rightarrow \infty} \frac{\sum_{j \in E_\nu} (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2} = 0.$$

Now, from (4.8) and (4.11) it follows that $E_{\nu i} \subset E_\nu$ and, consequently,

$$(4.14) \quad \frac{\sum_{i=1}^{N_\nu} \left[(b_{\nu i} - \bar{b}_\nu)^2 \sum_{j \in E_{\nu i}} (a_{\nu j} - \bar{a}_\nu)^2 \right]}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2} \leq \frac{\sum_{j \in E_\nu} (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2}.$$

Finally, (4.3) is an obvious consequence of (4.9), (4.14) and (4.13).

Second, let us assume that (2.10) does not hold. Then there exists a sequence B_ν of sets of integers such that, first, $B_\nu \subset \{1, \dots, N_\nu\}$, second, the numbers of elements in B_ν , say l_ν , satisfy the relation

$$(4.15) \quad \lim_{\nu \rightarrow \infty} (l_\nu / N_\nu) = 0,$$

and, third,

$$(4.16) \quad \limsup_{\nu \rightarrow \infty} \frac{\sum_{j \in B_\nu} (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2} > 0.$$

If

$$(4.17) \quad C_\nu = \left\{ j : (a_{\nu j} - \bar{a}_\nu)^2 > \frac{(N_\nu/l_\nu)^{\frac{1}{2}}}{N_\nu} \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \right\},$$

and $C_\nu^* = \{1, 2, \dots, N_\nu\} - C_\nu$ then, clearly,

$$\frac{\sum_{j \in B_\nu \cap C_\nu^*} (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} \leq \left(\frac{N_\nu}{l_\nu} \right)^{\frac{1}{2}} \frac{l_\nu}{N_\nu} = \left(\frac{l_\nu}{N_\nu} \right)^{\frac{1}{2}}$$

and, therefore, according to (4.15),

$$\lim_{\nu \rightarrow \infty} \frac{\sum_{j \in B_\nu \cap C_\nu^*} (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} = 0.$$

Consequently, in view of (4.16),

$$(4.18) \quad \limsup_{\nu \rightarrow \infty} \frac{\sum_{j \in C_\nu} (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} \geq \limsup_{\nu \rightarrow \infty} \frac{\sum_{j \in B_\nu} (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} > 0.$$

Now, put

$$(4.19) \quad b_{\nu 1} = \dots = b_{\nu n_\nu} = 1, \quad b_{\nu n_\nu + 1} = \dots = b_{\nu N_\nu} = 0,$$

where n_ν is determined by

$$(4.20) \quad n_\nu \leq (N_\nu/l_\nu)^{\frac{1}{2}} < n_\nu + 1 \quad (\nu \geq 1).$$

We have, obviously,

$$(4.21) \quad \lim_{\nu \rightarrow \infty} (n_\nu/N_\nu) = 0$$

and, in view of (4.15),

$$(4.22) \quad \lim_{\nu \rightarrow \infty} n_\nu = \infty$$

Furthermore,

$$(4.23) \quad \bar{b}_\nu = n_\nu/N_\nu,$$

$$(4.24) \quad \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 = [n_\nu(N_\nu - n_\nu)]/N_\nu,$$

and

$$(4.25) \quad \frac{\max_{1 \leq i \leq N_\nu} (b_{\nu i} - \bar{b}_\nu)^2}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2} = \frac{\max \left\{ \left(1 - \frac{n_\nu}{N_\nu} \right)^2, \left(\frac{n_\nu}{N_\nu} \right)^2 \right\}}{\frac{n_\nu(N_\nu - n_\nu)}{N_\nu}} \leq \max \left\{ \frac{1}{n_\nu}, \frac{1}{N_\nu - n_\nu} \right\}.$$

Consequently, the relations (4.21) and (4.22) ensure that the condition (4.2) is satisfied.

The sets $E_{\nu i}$, given generally by (4.8), will be now for $1 \leq i \leq n_\nu$ defined by

$$(4.26) \quad E_{\nu i} = E_{\nu 1} = \left\{ j : (a_{\nu j} - \bar{a}_\nu)^2 > \tau^2 \frac{n_\nu}{N_\nu - n_\nu} \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \right\} \quad (1 \leq i \leq n_\nu).$$

From (4.20) it follows that, for ν sufficiently large, the set C_ν given by (4.17) will be included in the set $E_{\nu 1}$ given by (4.26). Therefore

$$(4.27) \quad \begin{aligned} \frac{1}{N_\nu} \sum_{i=1}^{n_\nu} \sum_{|b_{\nu i j}| > \tau} \delta_{\nu i j}^2 &= \frac{\sum_{i=1}^{N_\nu} [(b_{\nu i} - \bar{b}_\nu)^2 \sum_{j \in E_{\nu i}} (a_{\nu j} - \bar{a}_\nu)^2]}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2} \\ &\geq \frac{\sum_{i=1}^{n_\nu} [(b_{\nu i} - \bar{b}_\nu)^2 \sum_{j \in E_{\nu i}} (a_{\nu j} - \bar{a}_\nu)^2]}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2} = \frac{\left(1 - \frac{n_\nu}{N_\nu}\right)^2 n_\nu \sum_{j \in E_{\nu 1}} (a_{\nu j} - \bar{a}_\nu)^2}{n_\nu \left(1 - \frac{n_\nu}{N_\nu}\right) \sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2} \\ &\geq \left(1 - \frac{n_\nu}{N_\nu}\right) \frac{\sum_{j \in C_\nu} (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{j=1}^{N_\nu} (a_{\nu j} - \bar{a}_\nu)^2} \quad (\nu \geq \nu_0). \end{aligned}$$

The relations (4.27), (4.18) and (4.21) imply that (4.3) cannot hold, and the theorem is thereby proved.

Let $G_\nu(x)$ denote the distribution function of the numbers $a_{\nu 1} \leq \dots \leq a_{\nu N_\nu}$, i.e.,

$$(4.28) \quad G_\nu(x) = \frac{\text{number of the } a\text{'s smaller or equal to } x}{N_\nu}.$$

LEMMA 4.1. *Condition (4.10) may be expressed in either of the following three forms:*

- (i) *The functions $[a_\nu(\lambda) - \bar{a}_\nu]^2 \{ \int_0^1 [a_\nu(\lambda) - \bar{a}_\nu]^2 d\lambda \}^{-1}$ are uniformly integrable.*
- (ii)

$$(4.29) \quad [\lim_{\nu \rightarrow \infty} \epsilon_\nu = 0] \Rightarrow \left[\lim_{\nu \rightarrow \infty} \frac{\left(\int_0^{\epsilon_\nu} + \int_{1-\epsilon_\nu}^1 \right) [a_\nu(\lambda) - \bar{a}_\nu]^2 d\lambda}{\int_0^1 [a_\nu(\lambda) - \bar{a}_\nu]^2 d\lambda} = 0 \right].$$

(iii)

$$(4.30) \quad [\lim_{\nu \rightarrow \infty} K_\nu = \infty] \Rightarrow \left[\lim_{\nu \rightarrow \infty} \frac{1}{\sigma_\nu^2} \int_{|x - \bar{a}_\nu| > K_\nu \sigma_\nu} (x - \bar{a}_\nu)^2 dG_\nu(x) = 0 \right],$$

where

$$(4.31) \quad \sigma_\nu^2 = \frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 = \int_0^1 [a_\nu(\lambda) - \bar{a}_\nu]^2 d\lambda = \int_{-\infty}^{\infty} (x - \bar{a}_\nu)^2 dG_\nu(x).$$

PROOF. If $a_{\nu 1} \leq \dots \leq a_{\nu N_\nu}$, then surely

$$(4.32) \quad \max_{1 \leq i_1 < \dots < i_{k_\nu} \leq N_\nu} \sum_{\alpha=1}^{k_\nu} (a_{\nu i_\alpha} - \bar{a}_\nu)^2 = \sum_{i=1}^{k_{\nu 1}} (a_{\nu i} - \bar{a}_\nu)^2 + \sum_{i=N_\nu - k_{\nu 2} + 1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2,$$

where $k_{\nu 1} + k_{\nu 2} = k_\nu$. On the other hand, we have

$$(4.33) \quad \sum_{i=1}^{k_{\nu 1}} (a_{\nu i} - \bar{a}_\nu)^2 + \sum_{i=N_\nu - k_{\nu 2} + 1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 = N_\nu \left(\int_0^{k_{\nu 1}/N_\nu} + \int_{1-(k_{\nu 2}/N_\nu)}^1 \right) [a_\nu(\lambda) - \bar{a}_\nu]^2 d\lambda,$$

which proves the equivalency of (4.10) to (ii) and, of course, to (i) as well.

Now we shall prove the equivalency of (4.10) to (4.30). Clearly,

$$(4.34) \quad \frac{1}{\sigma_\nu^2} \int_{|x - \bar{a}_\nu| > K_\nu \sigma_\nu} (x - \bar{a}_\nu)^2 dG_\nu(x) = \frac{1}{\sigma_\nu^2} \frac{1}{N_\nu} \sum_{|a_{\nu i} - \bar{a}_\nu| > K_\nu \sigma_\nu} (a_{\nu i} - \bar{a}_\nu)^2.$$

Denoting the number of a 's such that $|a_{\nu i} - \bar{a}_\nu| > K_\nu \sigma_\nu$ by k_ν , we have the following form of Tchebychev's inequality:

$$(4.35) \quad k_\nu K_\nu^2 \leq 1/\sigma_\nu^2 \sum_{|a_{\nu i} - \bar{a}_\nu| > K_\nu \sigma_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \leq N_\nu.$$

Assume, first, that (4.10) holds. Then $K_\nu \rightarrow \infty$ implies, in view of (4.35), $k_\nu/N_\nu \rightarrow 0$, so that, according to (4.10), the right side of (4.34) tends to 0, and, consequently, also the left side of (4.34) tends to 0. Thus (4.10) implies (4.30).

Assume, second, that (4.10) does not hold. On repeating the respective part of the proof of Theorem 4.2, we get again the relation (4.18), which is equivalent to

$$(4.36) \quad \limsup_{\nu \rightarrow \infty} \frac{1}{\sigma_\nu^2} \int_{|x - \bar{a}_\nu| > (N_\nu/l_\nu)^{\frac{1}{2}} \sigma_\nu} (x - \bar{a}_\nu)^2 dG_\nu(x) > 0.$$

This means that (4.30) is not satisfied for $K_\nu = (N_\nu/l_\nu)^{\frac{1}{2}}$. So the negation of (4.10) implies the negation of (4.30), i.e., (4.30) implies (4.10). Lemma 4.1 is proved.

COROLLARY. *The statistic (3.13) is asymptotically normal with mean value ES_ν and variance $\text{var } S_\nu$ for any double sequence $\{b_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ satisfying (4.2).*

PROOF. By Lemma 2.2 and Lemma 4.1 the numbers $a_{\nu i} = \varphi[i/(N_\nu + 1)]$ fulfill the condition (4.10). It suffices to apply Theorem 4.2.

EXAMPLE 4.1. If the b 's are given by (4.19), then the statistic

$$(4.37) \quad S_\nu = \sum_{i=1}^{N_\nu} b_{\nu i} a_{\nu R_{\nu i}} = \sum_{i=1}^{n_\nu} a_{\nu R_{\nu i}}$$

represents a sum of n_ν elements selected by simple random sampling from the population $\{a_{\nu 1}, \dots, a_{\nu N_\nu}\}$. The condition (4.2) is fulfilled, according to (4.25),

if, and only if,

$$(4.38) \quad \lim_{\nu \rightarrow \infty} n_\nu = \lim_{\nu \rightarrow \infty} (N_\nu - n_\nu) = \infty.$$

Hence, provided that (4.1) holds, the distribution of S_ν is asymptotically normal with mean value ES , and variance $\text{var } S_\nu$ for all n_ν satisfying (4.38) if, and only if, the a 's satisfy the condition (4.10).

As we shall see in Section 5, (4.10) is fulfilled, for example, if the populations $\{a_{\nu 1}, \dots, a_{\nu N_\nu}\}$ have uniformly bounded excesses. See also [9] and [10] for further results concerning sampling from a finite population.

5. Comparison of various conditions. First let us introduce the following

NOTATION. The condition (4.1), introduced by Noether [3] and simplified by Hoeffding [4], will be denoted by N .

The Lindeberg condition (4.3) will be denoted by L .

The condition (4.10) will be denoted by Q .

The Wald-Wolfowitz [1] condition

$$(5.1) \quad \frac{\frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^r}{\left[\frac{1}{N_\nu} \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \right]^{r/2}} = O(1) \quad (r = 3, 4, \dots)$$

where $O(1)$ denotes uniform boundedness, will be denoted by W .

The Hoeffding [4] condition

$$(5.2) \quad \lim_{\nu \rightarrow \infty} N_\nu^{\frac{1}{2}r-1} \frac{\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^r \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^r}{\left[\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2 \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \right]^{r/2}} = O$$

will be denoted by H .

Observe that the conditions L and H concern $\{a_{\nu i}, b_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ whereas the conditions N, Q, W are applied to each double sequence $\{a_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ and $\{b_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ separately. The fact that $\{b_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ satisfies N and $\{a_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ satisfies Q will be denoted by NQ , and the symbols NN, NW and WW will have similar interpretations.

THEOREM 5.1. $WW \Rightarrow NW \Rightarrow H, \quad NQ \Rightarrow L \Rightarrow NN$.

PROOF. For $WW \Rightarrow NW \Rightarrow H$ see Hoeffding [4], and for $H \Rightarrow L$ Motoo [8]. $NQ \Rightarrow L$ follows from Theorem 4.2. Thus it remains to prove $NW \Rightarrow NQ$, i.e., $W \Rightarrow Q$, and $L \Rightarrow NN$.

$W \Rightarrow Q$. If we take $r = 4$ in (5.1) and use the quantile function form, we get

$$(5.3) \quad \frac{\int_0^1 [a_\nu(\lambda) - \bar{a}_\nu]^4 d\lambda}{\left(\int_0^1 [a_\nu(\lambda) - \bar{a}_\nu]^2 d\lambda \right)^2} = O(1)$$

As is well-known from a theorem due to Vallé-Poussin, (5.3) implies that the functions $[a_\nu(\lambda) - \bar{a}_\nu]^2 / \int_0^1 [a_\nu(\lambda) - \bar{a}_\nu] d\lambda$ are uniformly integrable, which is equivalent to Q (see Lemma 4.1).

$L \Rightarrow NN$. This fact follows from the inequality

$$(5.4) \quad \frac{\max_{1 \leq j \leq N_\nu} (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} = \frac{\sum_{i=1}^{N_\nu} \max_{1 \leq j \leq N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 (a_{\nu j} - \bar{a}_\nu)^2}{\sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2 \sum_{i=1}^{N_\nu} (a_{\nu i} - \bar{a}_\nu)^2} \\ = \frac{1}{N_\nu} \sum_{i=1}^{N_\nu} \max_{1 \leq j \leq N_\nu} \delta_{\nu ij}^2 \leq \epsilon + \frac{1}{N_\nu} \sum_{|\delta_{\nu ij}| > \epsilon} \delta_{\nu ij}^2 \quad (\epsilon > 0),$$

where the a 's can be replaced by the b 's.

REMARK 5.1. The condition (5.3) means that the excesses are uniformly bounded. Denoting this condition by W_4 , we have $W \Rightarrow W_4 \Rightarrow Q$.

In [7] Dwass considered the empirical distribution function $G_\nu(x)$ of the values $a_{\nu 1}, \dots, a_{\nu N_\nu}$ and supposed that, first, $\lim_{\nu \rightarrow \infty} G_\nu(x) = G(x)$ at every continuity point of a distribution function $G(x)$ and, second,

$$(5.5) \quad \int x dG_\nu(x) = \int x dG(x) = 0$$

$$(5.6) \quad \int x^2 dG_\nu(x) = \int x^2 dG(x) = 1.$$

These assumptions imply that

$$[\lim_{\nu \rightarrow \infty} K_\nu = \infty] \Rightarrow \left[\lim_{\nu \rightarrow \infty} \int_{|x| > K_\nu} x^2 dG_\nu = 0 \right].$$

Consequently, by Lemma 4.1, the a 's fulfill the condition Q , so that the respective part of Dwass theorem [7] is contained in Theorem 4.2.

6. A special case encountered in rank order test theory. In rank order tests theory there are used locally most powerful tests based on statistics of the form

$$(6.1) \quad S_\nu = \sum_{i=1}^{N_\nu} b_{\nu i} E\{\varphi(U_i) \mid R_{\nu i}\}$$

where $E\{\cdot \mid R_{\nu i}\}$ denotes the conditional mean value under the condition that the rank of U_i among the observations U_1, \dots, U_{N_ν} equals $R_{\nu i}$. Let us observe that (6.1) is a special case of (1.1) for

$$(6.2) \quad a_{\nu i} = E\{\varphi(U_1) \mid R_{\nu 1} = i\} = E\{\varphi(Z_{\nu i})\},$$

where, in the middle expression, the index 1 might be replaced by any index $j = 1, \dots, N_\nu$.

For simplicity we shall suppose that, as in previous sections, the U 's have a uniform distribution. This causes no loss of generality, since arbitrarily distributed observations may be expressed as (non-decreasing) functions of uni-

formly distributed observations. We shall also assume that

$$(6.3) \quad \int_0^1 \varphi(\lambda) d\lambda = 0$$

and that

$$(6.4) \quad \int_0^1 \varphi^2(\lambda) d\lambda < \infty.$$

The assumption (6.3), however, is not essential. The integers N_ν will be assumed to tend to ∞ monotonically, i.e., $N_\nu \leq N_{\nu+1}$, $\nu \geq 1$.

LEMMA 6.1. *Put*

$$(6.5) \quad Y_\nu = Y_\nu(R_{\nu 1}) = E\{\varphi(U_1) | R_{\nu 1}\}$$

and assume that (6.3) and (6.4) hold. Then

(i)

$$(6.6) \quad P\{\lim_{\nu \rightarrow \infty} Y_\nu = \varphi(U_1)\} = 1,$$

(ii) $\{Y_1, Y_2, \dots, \varphi(U_1)\}$ is a martingale and $\{Y_1^2, Y_2^2, \dots, \varphi^2(U_1)\}$ is a semimartingale,

(iii) the random variables $Y_\nu^2 (\nu \geq 1)$ are uniformly integrable,

(iv)

$$(6.7) \quad \lim_{\nu \rightarrow \infty} E|Y_\nu^2 - \varphi^2(U_1)| = 0$$

and

$$(6.8) \quad \lim_{\nu \rightarrow \infty} EY_\nu^2 = E\varphi^2(U_1).$$

PROOF. The Borel fields \mathfrak{F}_ν generated by the random vectors $(R_{\nu 1}, \dots, R_{\nu N_\nu})$ form an increasing sequence of Borel fields. Denote by \mathfrak{F}_∞ the smallest Borel field containing $\bigcup_1^\infty \mathfrak{F}_\nu$. As is well-known, the conditional distribution of U_1 for given $R_{\nu 1} = j$ is the Beta distribution with $p = j$ and $q = N - j + 1$. Hence

$$(6.9) \quad E\left(U_1 - \frac{R_{\nu 1}}{N_{\nu+1}}\right)^2 = \frac{1}{N_\nu} \sum_{j=1}^{N_\nu} \frac{j(N_\nu - j + 1)}{(N_\nu + 1)^2(N_\nu + 2)} < \frac{1}{N_\nu},$$

from which follows that U_1 is equivalent (with probability 1) to a random variable measurable with respect to \mathfrak{F}_∞ . Consequently, $\varphi(U_1)$ is also equivalent to a random variable measurable with respect to \mathfrak{F}_∞ .

Since the conditional distribution of U_1 for fixed $R_{\nu 1}$ is independent of $R_{\nu 2}, \dots, R_{\nu N_\nu}$, we may also write

$$(6.10) \quad Y_\nu = E\{\varphi(U_1) | R_{\nu 1}, \dots, R_{\nu N_\nu}\} = E\{\varphi(U_1) | \mathfrak{F}_\nu\}.$$

Now we can apply the theory of martingales. The assertions (i) through (iv) are consequences of Doob [13], Chap. VII, Theorem 4.4, §1 Example 1, Theorem

1.1, Theorem 3.3 (III), Theorem 4.1 s, respectively. (6.8) is a simple consequence of (6.7).

LEMMA 6.2. *The numbers $a_{\nu i}$ given by (6.2) satisfy the condition Q given by (4.10).*

PROOF. Observe that Y_ν takes on the values $a_{\nu i}$ with equal probabilities so that condition (4.10) coincides, in view of (6.3) and (6.8), with uniform integrability of the random variables Y_ν , $\nu \geq 1$, which has been proved by Lemma 6.1.

THEOREM 6.1. *If $\{b_{\nu i}, 1 \leq i \leq N_\nu, \nu \geq 1\}$ fulfill the condition N given by (4.2) and the function $\varphi(\lambda)$ is non-vanishing and quadratically integrable, then the statistic (6.1) has an asymptotically normal distribution with mean value ES_ν and variance $\text{var } S_\nu$.*

PROOF. It suffices to apply Theorem 4.2 and Lemma 6.2.

THEOREM 6.2. *Under the assumptions (6.3) and (6.4), the statistic (6.1) is asymptotically equivalent in the mean to the statistic*

$$(6.11) \quad T_\nu = \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu) \varphi(U_i).$$

PROOF. By the method used in proving (3.9), we can show that

$$(6.12) \quad E(S_\nu - T_\nu)^2 \leq [N_\nu / (N_\nu - 1)] E[\varphi(U_1) - Y_\nu]^2 \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2.$$

In view of (6.5), (6.12) is equivalent to

$$(6.13) \quad E(S_\nu - T_\nu)^2 \leq [N_\nu / (N_\nu - 1)] [E\varphi^2(U_1) - EY_\nu^2] \sum_{i=1}^{N_\nu} (b_{\nu i} - \bar{b}_\nu)^2.$$

Now it only remains to divide both sides of (6.13) by $\text{var } T_\nu$ and to apply (6.8).

REMARK 6.1. Theorem 6.1 generalizes the Dwass theorem [6], and the equality (6.8) generalizes the respective part of Hoeffding's Theorem 2 [5]. The condition of convexity of $\varphi(\lambda)$ is removed, which proves the conjecture made by Dwass in [14], p. 358.

7. Vector considerations. We shall briefly touch the question of asymptotical m -dimensional normality of a vector $(S_\nu^1, \dots, S_\nu^m)$ where

$$(7.1) \quad S_\nu^g = \sum_{i=1}^{N_\nu} b_{\nu i} a_{\nu R_{\nu i}}^g \quad (g = 1, \dots, m),$$

i.e., the b 's are fixed and the a 's depend on $g = 1, \dots, m$.

THEOREM 7.1. *Suppose that for any constants $\lambda_1, \dots, \lambda_m$*

$$(7.2) \quad \sum_{i=1}^{N_\nu} \left[\sum_{g=1}^m \lambda_g (a_{\nu i}^g - \bar{a}_\nu^g) \right]^2 \geq \epsilon \max_{1 \leq g \leq m} \left[\lambda_g^2 \sum_{i=1}^{N_\nu} (a_{\nu i}^g - \bar{a}_\nu^g)^2 \right]$$

where ϵ is positive and independent of $\nu \geq 1$. Assume that the b 's fulfill the condition N given by (4.2) and the a^g 's ($g = 1, \dots, m$) the condition Q given by (4.10). Then the vector $(S_\nu^1, \dots, S_\nu^m)$ has an asymptotically normal distribution with mean values ES_ν^g and covariances $\text{cov}(S_\nu^g, S_\nu^h)$, $1 \leq g, h \leq N_\nu$, $\nu \geq 1$.

PROOF. According to a theorem by Cramér [12], p. 105, it suffices to prove asymptotic normality for any linear combination

$$\sum_{g=1}^m \lambda_g S_{\nu}^g = \sum_{i=1}^{N_{\nu}} b_{\nu i} \left[\sum_{g=1}^m \lambda_g a_{\nu R_{\nu} i}^g \right].$$

This will be done if we show that the numbers $c_{\nu i} = \sum_{g=1}^m \lambda_g a_{\nu i}^g$ fulfill the condition (4.10) for any $\lambda_1, \dots, \lambda_m$. However, in view of (7.2), we have

$$\begin{aligned} \frac{\sum_{\alpha=1}^{k_{\nu}} (c_{\nu i_{\alpha}} - \bar{c}_{\nu})^2}{\sum_{i=1}^{N_{\nu}} (c_{\nu i} - \bar{c}_{\nu})^2} &= \frac{\sum_{\alpha=1}^{k_{\nu}} \left[\sum_{g=1}^m \lambda_g (a_{\nu i_{\alpha}}^g - \bar{a}_{\nu}^g) \right]^2}{\sum_{i=1}^{N_{\nu}} \left[\sum_{g=1}^m \lambda_g (a_{\nu i}^g - \bar{a}_{\nu}^g) \right]^2} \\ &\leq \frac{\sum_{\alpha=1}^{k_{\nu}} m \sum_{g=1}^m \lambda_g^2 (a_{\nu i_{\alpha}}^g - \bar{a}_{\nu}^g)^2}{\epsilon \max_{1 \leq g \leq m} \left[\lambda_g^2 \sum_{i=1}^{N_{\nu}} (a_{\nu i}^g - \bar{a}_{\nu}^g)^2 \right]} \leq \frac{m}{\epsilon} \frac{\sum_{\alpha=1}^{k_{\nu}} \sum_{g=1}^m (a_{\nu i_{\alpha}}^g - \bar{a}_{\nu}^g)^2}{\sum_{i=1}^{N_{\nu}} \sum_{g=1}^m (a_{\nu i}^g - \bar{a}_{\nu}^g)^2}. \end{aligned}$$

Now it suffices to note that, according to our suppositions the values $a_{\nu i}^g$ fulfill the condition (4.10).

REMARK 7.1. The condition (7.2.) simply means that all multiple correlation coefficients are uniformly bounded from 1.

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