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INTERACTIONS IN FACTORIAL EXPERIMENTS

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# ON A CLASS OF NONPARAMETRIC TESTS FOR INTERACTIONS IN FACTORIAL EXPERIMENTS\*

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1. Summary and introduction. This paper deals with a class of permutationally distribution-free aligned rank order tests for interactions in factorial experiments replicated in complete blocks. The asymptotic power-efficiencies of the proposed tests with respect to the classical analysis of variance test are also studied.

Nonparametric analysis of variance tests, available in the literature, mostly relate to one way or two way (without interaction) layouts. Though the approach of Lehmann (1964) (see also Puri and Sen (1966)) can be adapted to construct tests for interactions in factorial experiments, the necessity of avoiding incompatibility of the unadjusted estimates as well as of estimating some functional of the parent distribution ( appearing in the expression for the dispersion matrix of the estimators) makes such tests only asymptotically distribution-free and somewhat tedious to apply. In the present paper, the theory of aligned rank order tests based on Chernoff-Savage (1958) type of statistics, developed in Sen (1966b), is further extended to provide suitable tests for interactions in factorial layouts with equal number of observations per cell. Under certain permutational invariance arguments the nonparametric structure of the proposed tests is established. These tests are also free from the other two difficulties mentioned earlier. Further, using a generalization of Chernoff-Savage (1958) theorem on the asymptotic normality of rank order statistics to aligned observations, the asymptotic power-efficiencies of the proposed tests (along with certain bounds are studied.

2. Preliminary notions. We shall consider in detail only the case of replicated two factor experiments with one observation per call and indicate briefly the theory

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for the case of several factors and/or observations per cell. The chance variable  $Y_{ijk}$  associated with the yield of the plot placed in the  $i$ th replicate and receiving the combination of the  $j$ th variety (or level of the first factor) and the  $k$ -th treatment (or level of the second factor), is expressed, in accordance with the usual fixed-effects model, as

$$(2.1) \quad Y_{ijk} = \mu_i + v_j + \tau_k + \eta_{jk} + U_{ijk}, \quad i=1, \dots, n; \quad j=1, \dots, p; \quad k=1, \dots, q,$$

where  $\mu_1, \dots, \mu_n$  stand for the replication-effects,  $v_1, \dots, v_p$  for the variety-effects,  $\tau_1, \dots, \tau_q$  for the treatment-effects,  $\eta_{11}, \dots, \eta_{pq}$  for the variety x treatment interactions, and  $U_{ijk}$ 's are the residual error components. In (2.1), we may put

$$(2.2) \quad \sum_{j=1}^p v_j = 0, \quad \sum_{k=1}^q \tau_k = 0, \quad \sum_{k=1}^q \eta_{jk} = 0, \quad j=1, \dots, p, \quad \text{and} \quad \sum_{j=1}^p \eta_{jk} = 0, \quad k=1, \dots, q.$$

It is assumed that  $(U_{i11}, \dots, U_{ipq})$ ,  $i=1, \dots, n$  are independent and identically distributed stochastic vectors having a common continuous (joint) cumulative distribution function (cdf)  $F(x_{11}, \dots, x_{pq})$  which is symmetric in its  $pq$  arguments; this includes the conventional assumption of independence and identity of distributions of all the  $npq$  error components as a particular case. We want to test

$$(2.3) \quad H_0: \quad \underline{\eta} = (\eta_{jk}) = \underline{0}^{p \times q},$$

against the set of alternatives that  $\underline{\eta}$  is non-null. By means of the following intra-block transformations, we eliminate the nuisance parameters  $v_j$ 's and  $\tau_k$ 's. Let us consider the  $p \times q$  matrices

$$(2.4) \quad \underline{Y}_i = (Y_{ijk}), \quad \underline{U}_i = (U_{ijk}), \quad \underline{Z}_i = (Z_{ijk}) \quad \text{and} \quad \underline{E}_i = (E_{ijk}), \quad i=1, \dots, n,$$

where we define

$$(2.5) \quad Z_i = \left( I_p - \frac{1}{p} \mathbf{l}_p \mathbf{l}_p' \right) Y_i \left( I_q - \frac{1}{q} \mathbf{l}_q \mathbf{l}_q' \right), \quad i=1, \dots, n;$$

$$(2.6) \quad E_i = \left( I_p - \frac{1}{p} \mathbf{l}_p \mathbf{l}_p' \right) U_i \left( I_q - \frac{1}{q} \mathbf{l}_q \mathbf{l}_q' \right), \quad i=1, \dots, n,$$

$I_t$  being the identity matrix of order  $t$  and  $\mathbf{l}_t$  the (row)  $t$ -vector having all the  $t$  elements equal to unity,  $t \geq 1$ . Then from (2.1) through (2.6), we obtain that

$$(2.7) \quad Z_i = \eta + E_i, \quad i=1, \dots, n.$$

In the sequel, we shall work with the nuisance parameter-free model (2.7). Also, we will only consider the case when  $p, q \geq 3$ . If either of them is less than 3, the situation simplifies as follows. Suppose  $q=2, p > 2$ , then from (2.1) and (2.2) we have  $\eta_{j1} = -\eta_{j2} = \eta_j$  (say),  $j=1, \dots, p$ ; thus (2.3) reduces to  $H_0^*$ :  $\eta_1 = \dots = \eta_p = 0$ . Again from (2.5) and (2.6), we have

$$(2.8) \quad Z_{ij1} = -Z_{ij2} = Z_{ij} \text{ (say)}, \quad e_{ij1} = -e_{ij2} = e_{ij} \text{ (say)} \text{ for all } i=1, \dots, n.$$

It follows from lemma 3.1 [to be proved in section 3] that  $(e_{i1}, \dots, e_{ip})$  are symmetric dependent or interchangeable random variables for all  $i=1, \dots, n$ . Consequently, based on the set of observations  $\{Z_{ij}, j=1, \dots, p; i=1, \dots, n\}$ , the problem of testing  $H_0$  in (2.3) reduces to that of testing the interchangeability of  $(Z_{i1}, \dots, Z_{ip})$  (for all  $i=1, \dots, n$ ), against shift alternatives. As such, the results of Sen (1966b) will directly apply, and the details are omitted. If  $p=q=2$ , we have  $\eta_{11} = -\eta_{12} = -\eta_{21} = \eta_{22} = \eta$  (say), and

$$(2.9) \quad Z_{i11} = -Z_{i12} = -Z_{i21} = Z_{i22} = Z_i \text{ (say)}, \quad i=1, \dots, n,$$

and it is also easily seen that the distribution of  $Z_i$  is symmetric about  $\eta$ . So the problem of testing  $H_0$  in (2.3) reduces to that of testing the symmetry (about zero)

of the distribution of  $Z_i$ ; this is the well known one sample location problem, and hence, is not discussed.

3. The basic permutation principle. We define  $\underline{U}_i$  and  $\underline{E}_i$  ( $i=1, \dots, n$ ) as in (2.4) and (2.6), and let  $F^*(\underline{E})$  be the joint cdf of  $E_i$  for  $i=1, \dots, n$ . Let  $\underline{j} = (j_1, \dots, j_p)$  be any permutation of  $(1, \dots, p)$  and  $\underline{J}$ , the set of all possible ( $p!$ ) permutations, so that  $\underline{j} \in \underline{J}$ . Also let

$$(3.1) \quad \underline{I}_{\underline{p}}(\underline{j}) = (\delta_{i j_k})_{i,k=1, \dots, p}$$

where  $\delta_{rs}$  is the usual Kronecker delta. Now for any  $\underline{j} \in \underline{J}$ ,  $\underline{I}_{\underline{p}}(\underline{j})$  is non-singular and has a unique reciprocal  $\underline{I}_{\underline{p}}(\underline{j}^*)$  (say), which also belongs to the set

$\{\underline{I}_{\underline{p}}(\underline{j}) : \underline{j} \in \underline{J}\}$ . Further, it is easily seen that if  $\underline{I}_{\underline{p}}(\underline{j}_1)$  and  $\underline{I}_{\underline{p}}(\underline{j}_2)$  be defined as in (3.1) for  $\underline{j}_1 \in \underline{J}$ ,  $\underline{j}_2 \in \underline{J}$ , then  $\underline{I}_{\underline{p}}(\underline{j}_1)\underline{I}_{\underline{p}}(\underline{j}_2)$  also belongs to  $\{\underline{I}_{\underline{p}}(\underline{j}) : \underline{j} \in \underline{J}\}$ . Thus the set  $\{\underline{I}_{\underline{p}}(\underline{j}) : \underline{j} \in \underline{J}\}$  forms a finite group of elementary transformations.

It can also be verified that

$$(3.2) \quad \underline{I}_{\underline{p}}(\underline{j})\left(\underline{I}_{\underline{p}} - \frac{1}{p} \sum_{\underline{l} \in \underline{J}} \underline{I}_{\underline{p}}(\underline{l})\right)\underline{I}_{\underline{p}}(\underline{j}^*) = \underline{I}_{\underline{p}} - \frac{1}{p} \sum_{\underline{l} \in \underline{J}} \underline{I}_{\underline{p}}(\underline{l}) \quad \text{if} \quad \underline{I}_{\underline{p}}(\underline{j})\underline{I}_{\underline{p}}(\underline{j}^*) = \underline{I}_{\underline{p}}.$$

Similarly, let  $\underline{k} = (k_1, \dots, k_q)$  be any permutation of  $(1, \dots, q)$ ; the set of all possible ( $q!$ ) permutations is denoted by  $\underline{K}$ . A second group of elementary transformation matrices is then defined by

$$(3.3) \quad \{\underline{I}_{\underline{q}}(\underline{k}) : \underline{k} \in \underline{K}\} \quad \text{where} \quad \underline{I}_{\underline{q}}(\underline{k}) = (\delta_{i k_\ell})_{i,\ell=1, \dots, q}$$

Let us then define a finite group  $\mathcal{G}_i$  of transformations  $\{g_i(\underline{j}, \underline{k}) : \underline{j} \in \underline{J}, \underline{k} \in \underline{K}\}$  by

$$(3.4) \quad \begin{aligned} \underline{E}_i(\underline{j}, \underline{k}) &= g_i(\underline{j}, \underline{k})\underline{E}_i \\ &= \underline{I}_{\underline{p}}(\underline{j})\underline{E}_i\underline{I}_{\underline{q}}(\underline{k}), \quad \underline{j} \in \underline{J}, \underline{k} \in \underline{K}, \quad \text{for } i=1, \dots, n. \end{aligned}$$

Finally, the group of all the  $(p!q!)^n$  transformations in (3.4) is denoted by  $\mathcal{G}_n^*$  i.e.,

$$(3.5) \quad \mathcal{G}_n^* = (\mathcal{G}_1, \dots, \mathcal{G}_n).$$

As before, we denote the cdf's of  $U_i$  and  $E_i$  by  $F$  and  $F^*$ , respectively. Let now  $\mathcal{F}$  be the class of all pq-variate continuous cdf's for which the pq variates are interchangeable. By definition (in section 2),  $F \in \mathcal{F}$ .

LEMMA 3.1 If  $F \in \mathcal{F}$ ,  $F^*$  is  $\mathcal{G}$ -invariant.

PROOF. On defining  $U_i$ ,  $I_p(j)$  and  $I_q(k)$  as in (2.4), (3.1) and (3.3), we let

$$(3.6) \quad U_i(j, k) = I_p(j) U_i I_q(k), \quad j \in J, \quad k \in K.$$

Since  $F \in \mathcal{F}$ , it remains invariant under row (or column) permutations. Hence,  $U_i(j, k)$  has the same cdf  $F$  for all  $j \in J, k \in K$ . Now, from (2.6), (3.2) and (3.4), we obtain that

$$\begin{aligned} E_i(j, k) &= I_p(j) E_i I_q(k) \\ &= I_p(j) (I_p - \frac{1}{p} \sum_{p'} I_{p'}) U_i (I_q - \frac{1}{q} \sum_{q'} I_{q'}) I_q(k) \\ (3.7) \quad &= I_p(j) (I_p - \frac{1}{p} \sum_{p'} I_{p'}) I_p(j^*) I_p(j) U_i I_q(k) I_q(k^*) (I_q - \frac{1}{q} \sum_{q'} I_{q'}) I_q(k) \\ &= (I_p - \frac{1}{p} \sum_{p'} I_{p'}) U_i(j, k) (I_q - \frac{1}{q} \sum_{q'} I_{q'}) \end{aligned}$$

Thus, the invariance of the cdf of  $U_i(j, k)$  (under  $\mathcal{G}_i$ ) implies the invariance of the cdf  $E_i(j, k)$  under  $\mathcal{G}_i$ . Hence the lemma.

Let now  $Z_n^*$  be the npq-dimensional (Euclidean) space of the sample point  $Z_n^* = (Z_1, \dots, Z_n)$ . Then the finite group  $(\mathcal{G}_n^*)$  of transformations in (3.4) and (3.5) maps the sample space onto itself, and under  $H_0$  in (2.3), the distribution

of  $Z_n^*$  is  $\mathcal{G}_n^*$ -invariant. Thus, proceeding as in Hoeffding (1952, pp. 169-170), we may prove the existence of similar size  $\alpha$  ( $0 < \alpha < 1$ ) tests for  $H_0$  in (2.3), valid for all  $F \in \mathfrak{F}$ . These tests are essentially conditional tests based on the consideration of equiprobable all possible row and column permutations of the matrices  $Z_1, \dots, Z_n$ . Such a conditional test is termed a permutation test. In this paper, we shall study permutation tests based on a celebrated class of rank order statistics due to Chernoff and Savage (1958).

4. Formulation of the rank order tests. Let  $c(u)$  be 1,  $\frac{1}{2}$  or 0 according as  $u$  is  $>$ ,  $=$  or  $<$  0, and let

$$(4.1) \quad R_{ijk} = \frac{1}{2} + \sum_{r=1}^n \sum_{s=1}^p \sum_{t=1}^q c(Z_{ijk} - Z_{rst}), \quad i=1, \dots, n; \quad j=1, \dots, p; \quad k=1, \dots, q;$$

by virtue of the assumed continuity of  $F$ , ties among  $Z_{ijk}$ 's may be ignored, in probability. We define a sequence of real numbers  $\mathfrak{S}_N = (S_{N1}, \dots, S_{NN})$ , where

$$(4.2) \quad S_{N\alpha} = J_N(\alpha/(N+1)), \quad 1 \leq \alpha \leq N;$$

the function  $J_N$  is defined as in Chernoff and Savage (1958) and is assumed to satisfy the regularity conditions of theorem 1 of Chernoff and Savage (1958).

Let

$$(4.3) \quad \mathfrak{T}_N = (T_{N, jk}); \quad T_{N, jk} = \frac{1}{n} \sum_{i=1}^n S_{NR_{ijk}} \quad \text{for } j=1, \dots, p, \quad k=1, \dots, q.$$

Also, let

$$(4.4) \quad \mathfrak{T}_N^* = \left( \mathbf{I}_{np} - \frac{1}{p} \mathfrak{L}_{pp} \right) \mathfrak{T}_N \left( \mathbf{I}_q - \frac{1}{q} \mathfrak{L}_{qq} \right) = (T_{N, jk}^*),$$

$$(4.5) \quad S_{NR_{ijk}}^* = S_{NR_{ijk}} - \frac{1}{p} \sum_{j=1}^p S_{NR_{ijk}} - \frac{1}{q} \sum_{k=1}^q S_{NR_{ijk}} + \frac{1}{pq} \sum_{j=1}^p \sum_{k=1}^q S_{NR_{ijk}},$$

for  $i=1, \dots, n$ ;  $j=1, \dots, p$ ,  $k=1, \dots, q$ . Then, from (4.3), (4.4) and (4.5), we obtain

$$(4.6) \quad T_{N, jk}^* = \frac{1}{n} \sum_{i=1}^n S_{NR_{ijk}}^*, \quad j=1, \dots, p, \quad k=1, \dots, q.$$

We denote by  $\mathcal{P}_N$ , the permutational (conditional) probability measure induced by the  $(p!q!)^n$  equally likely transformations in  $\mathcal{F}_n^*$ , defined by (3.4) and (3.5).

Then, by simple arguments it follows that

$$(4.7) \quad E\{T_{N, jk}^* | \mathcal{P}_N\} = 0^{pxq}.$$

Also, let

$$(4.8) \quad \sigma^2(\mathcal{P}_N) = \frac{1}{n(p-1)(q-1)} \sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q (S_{NR_{ijk}}^*)^2,$$

and let  $\otimes$  stand for the symbol for Kronecker product of two square matrices. Then by routine computations, we obtain that

$$(4.9) \quad n E\{T_{N, jk}^* \otimes T_{N, jk}^* | \mathcal{P}_N\} = (I_{\sim p} - \frac{1}{p} \sum_{l=1}^p l_{\sim p}) \otimes (I_{\sim q} - \frac{1}{q} \sum_{l=1}^q l_{\sim q}) \sigma^2(\mathcal{P}_N).$$

Thus considering the generalized inverse of the  $pq \times pq$  matrix in (4.9) and employing it to construct a quadratic form in the elements of  $T_{N, jk}^*$ , we derive the following test statistic

$$(4.10) \quad \mathcal{L}_N = [n/\sigma^2(\mathcal{P}_N)] \sum_{j=1}^p \sum_{k=1}^q \{T_{N, jk}^*\}^2,$$



which is analogous to the classical parametric test based on the variance ratio criterion [cf. Scheffé (1959)].

For small values of  $n$ ,  $p$  and  $q$ , the exact permutation distribution of  $\mathcal{L}_N$  can be obtained by considering the  $(p!q!)^n$  (conditionally) equally likely row and column permutations of the matrices  $S_N^{(i)} = (S_{NR_{ijk}})$ ,  $i=1, \dots, n$ . This procedure becomes prohibitively laborious for large values of  $n$ ,  $p$  or  $q$ . For this reason, we consider the following large sample approach.

Let us denote the marginal cdf of  $Z_{ijk}$  by  $F_{[jk]}^*(x)$  and the joint cdf of  $(Z_{ijk}, Z_{ij'k'})$  by  $F_{[jk;j'k']}^*(x, y)$  for all  $j, j'=1, \dots, p$ ;  $k, k'=1, \dots, q$ . Let then

$$(4.11) \quad H(x) = \frac{1}{pq} \sum_{j=1}^p \sum_{k=1}^q F_{[jk]}^*(x);$$

$$(4.12) \quad H_{10}(x, y) = \frac{1}{qp(p-1)} \sum_{k=1}^q \sum_{j' \neq j=1}^p F_{[jk;j'k]}^*(x, y);$$

$$(4.13) \quad H_{01}(x, y) = \frac{1}{pq(q-1)} \sum_{j=1}^p \sum_{k' \neq k=1}^q F_{[jk;jk']}^*(x, y);$$

$$(4.14) \quad H_{11}(x, y) = \frac{1}{p(p-1)q(q-1)} \sum_{j' \neq j=1}^p \sum_{k' \neq k=1}^q F_{[jk;j'k']}^*(x, y).$$

We denote by  $J(u) = \lim_{N \rightarrow \infty} J_N(u)$ :  $0 < u < 1$ , and define

$$(4.15) \quad \delta^2 = \int_0^1 J^2(u) du - \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} J[H(x)]J[H(y)]d[H_{10}(x, y) + H_{01}(x, y) - H_{11}(x, y)].$$

Then, proceeding as in the proof of theorem 4.2 of Puri and Sen (1966) and omitting the details, we obtain the following theorem.

THEOREM 4.1 Under the conditions of theorem 1 of Chernoff and Savage (1958),

$[\sigma^2(\mathcal{P}_N) - \delta^2]$  converges in probability to zero.

Now, if we assume that

$$(4.16) \quad P\{[J[H(Z_{jk})] - J[H(Z_{j'k'})] - J[H(Z_{jk'})] + J[H(Z_{j'k'})]] = \text{Constant}\} < 1,$$

for at least one pair of  $j \neq j'$  and  $k \neq k'$ , then as in theorem 4.1 of Sen (1966b), it can be shown that  $\delta^2$ , defined by (4.15), is strictly positive. (4.16) will be termed, in the sequel, as the non-degeneracy condition of the cdf  $F^*$ . The main theorem of this section is the following.

THEOREM 4.2 Under the conditions of theorem 1 of Chernoff and Savage (1958), the permutation distribution of  $\mathcal{L}_N$  converges to a chi-square distribution with  $(p-1)(q-1)$  degrees of freedom (d.f.).

PROOF. By virtue of (4.7), (4.9) and (4.10), it suffices to show that for any non-null  $\mathbf{A} = (a_{jk})$ ,  $\mathbf{Y}_n = n^{-\frac{1}{2}} \sum_{j=1}^p \sum_{k=1}^q a_{jk} \mathbf{T}_{N, jk}^*$  converges in law (under  $\mathcal{P}_N$ ) to a normal distribution as  $n \rightarrow \infty$ . Now, using (4.4), (4.5), (4.6) and the first two conditions of theorem 1 of Chernoff and Savage (1958), we write

$$(4.17) \quad \mathbf{Y}_n = n^{-\frac{1}{2}} \sum_{i=1}^n \mathbf{Y}_{ni} + o_p(1); \mathbf{Y}_{ni} = \sum_{j=1}^p \sum_{k=1}^q a_{jk}^* J\left(\frac{R_{ijk}}{N+1}\right), \quad i=1, \dots, n,$$

where  $a_{jk}^*$ 's are linear functions of  $a_{jk}$ 's and they satisfy the constraints that  $\sum_{k=1}^q a_{jk}^* = 0$ ,  $j=1, \dots, p$  and  $\sum_{j=1}^p a_{jk}^* = 0$ ,  $k=1, \dots, q$ . We note that under  $\mathcal{P}_N$ ,  $\mathbf{Y}_{ni}$  can have only  $p!q!$  (conditionally) equally likely values obtained by permuting the rows and columns of the matrix  $\mathbf{R}_i = (R_{ijk})_{j=1, \dots, p, k=1, \dots, q}$ , and  $\mathbf{Y}_{n1}, \dots, \mathbf{Y}_{nn}$  are all stochastically independent (under  $\mathcal{P}_N$ ). Thus, it readily follows that

$$E(\mathbf{Y}_{ni} | \mathcal{P}_N) = 0, \text{ and}$$

$$(4.18) \quad E(Y_{ni}^2 | \mathcal{P}_N) = \sum_{j=1}^p \sum_{k=1}^q (a_{jk}^*)^2 \cdot \frac{1}{(p-1)(q-1)} \left\{ \sum_{j=1}^p \sum_{k=1}^q J^2\left(\frac{R_{ijk}}{N+1}\right) - \frac{1}{p} \sum_{k=1}^q \left( \sum_{j=1}^p J\left(\frac{R_{ijk}}{N+1}\right) \right)^2 - \frac{1}{q} \sum_{j=1}^p \left( \sum_{k=1}^q J\left(\frac{R_{ijk}}{N+1}\right) \right)^2 + \frac{1}{pq} \left( \sum_{j=1}^p \sum_{k=1}^q J\left(\frac{R_{ijk}}{N+1}\right) \right)^2 \right\} .$$

Thus, by routine analysis, it follows as in theorem 4.1 that

$$(4.19) \quad \frac{1}{n} \sum_{i=1}^n E(Y_{ni}^2 | \mathcal{P}_N) \rightarrow \delta^2 \sum_{j=1}^p \sum_{k=1}^q (a_{jk}^*)^2 > 0,$$

where  $\delta^2$  is defined by (4.15) and is positive by (4.16). Further, using the growth condition of theorem 1 of Chernoff and Savage (1958), it follows that

$$(4.20) \quad \frac{1}{n} \sum_{i=1}^n E(|Y_{ni}|^{2+\delta} | \mathcal{P}_N) < \infty, \text{ for some } \delta > 0.$$

Consequently, using the Berry-Esseen theorem (cf. [ 4 , p. 288]), the asymptotic normality of  $Y_n$  follows from (4.19) and (4.20). Hence the theorem.

By virtue of theorem 4.2, an asymptotically size  $\alpha$  ( $0 < \alpha < 1$ ) test for the hypothesis of no interaction may be proposed as follows. If

$$(4.21) \quad \mathcal{L}_N \begin{cases} \geq \chi_{(p-1)(q-1), \alpha}^2, & \text{reject } H_0 \text{ in (2.3),} \\ < \chi_{(p-1)(q-1), \alpha}^2, & \text{accept } H_0, \end{cases}$$

where  $\chi_{t, \alpha}^2$  is the upper 100 $\alpha$ % point of a chi-square distribution having  $t$  d.f. .

5. Asymptotic efficiency of the test based on  $\mathcal{L}_N$ . It can be easily shown that the test in (4.19) is consistent for any non-null  $\underline{\eta} = (\eta_{jk})$ . For the study of

the asymptotic efficiency of the test based on  $\mathcal{L}_N$ , we shall therefore consider the following sequence of Pitman-alternatives, specified by

$$(5.1) \quad H_N: \mathcal{T} = \mathcal{T}_N = N^{-\frac{1}{2}} \mathcal{A}, \quad \mathcal{A} = (\lambda_{jk}),$$

where  $\lambda_{jk}$ 's are real and finite and they satisfy

$$(5.2) \quad \sum_{j=1}^p \lambda_{jk} = 0, \quad k=1, \dots, q; \quad \sum_{k=1}^q \lambda_{jk} = 0 \quad \text{for } j=1, \dots, p.$$

Thus, under  $\{H_N\}$ , the cdf of  $Z_i$  (defined by (2.5),) is specified by  $F^*(\underline{x} - N^{-\frac{1}{2}}\mathcal{A})$ , (where  $\underline{x}$  is a  $p \times q$  matrix), and  $F^*(\underline{x})$  is invariant under the row or column permutations of  $\underline{x}$ . Thus, the univariate marginal cdf  $F_{[jk]}^*(x)$  (of  $Z_{ijk} - N^{-\frac{1}{2}}\lambda_{jk}$ ) is independent of  $(j, k)$  and is denoted by  $H(x)$  [cf. (4.11)]. Similarly, the bivariate marginal cdf  $F_{[jk, jk']}^*(x, y)$  will be independent of  $(j, k \neq k')$  and is denoted by  $H_{01}(x, y)$  [cf. (4.13)],  $F_{[jk, j'k]}^*(x, y)$  will be independent of  $(j \neq j', k)$  and is denoted by  $H_{10}(x, y)$  [cf. (4.12)], and  $F_{[jk; j'k']}^*(x, y)$  will be independent of  $(j \neq j', k \neq k')$  and is denoted by  $H_{11}(x, y)$  [cf. (4.14)].

Now, for arbitrary  $F^*(\underline{x})$ , the asymptotic normality of  $N^{\frac{1}{2}}(\mathcal{T}_N - \mu_N)$  (where  $\mu_N = (\mu_{N, jk})$ ,  $\mu_{N, jk} = \int_{-\infty}^{\infty} J[H(x)] dF_{[jk]}^*(x)$ ,  $j=1, \dots, p$ ,  $k=1, \dots, q$ ) can be proved along the same line as in the proof of theorem 5.1 of Sen (1966a) (with direct-extension to the matrix case). We shall specifically consider the case when  $\{H_N\}$  holds. For this, we define

$$(5.3) \quad B(H) = \int_{-\infty}^{\infty} \frac{d}{dx} J[H(x)] dH(x),$$

$$(5.4) \quad A^2 = \int_0^1 J^2(u) du - \left[ \int_0^1 J(u) du \right]^2,$$

$$(5.5) \quad \rho_{ij} = \frac{1}{A^2} \left[ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} J[H(x)]J[H(y)]dH_{ij}(x,y) - \left\{ \int_0^1 J(u)du \right\}^2 \right],$$

for  $(i, j) = (0, 1), (1, 0)$  and  $(1, 1)$ , where  $H_{ij}$ 's are defined earlier. Then, by the same technique as in theorem 5.1 of Sen (1966a), we obtain that under  $\{H_N\}$ ,

$$(5.6) \quad n^{\frac{1}{2}}E\{\underline{T}_N^*|H_N\} = [(pq)^{-\frac{1}{2}}B(H)] \underline{\Delta} + o(1)$$

$$(5.7) \quad n E\{\underline{T}_N^* \otimes \underline{T}_N^*|H_N\} = \left( \underline{I}_p - \frac{1}{p} \underline{\rho}_{pp} \right) \otimes \left( \underline{I}_q - \frac{1}{q} \underline{\rho}_{qq} \right) A^2(1-\rho_{10}-\rho_{01}+\rho_{11})+o(1),$$

where  $B(H)$ ,  $A^2$ ,  $\rho_{ij}$ 's are defined in (5.3), (5.4) and (5.5). Again, using (4.15), theorem 4.1 and some routine computations, it follows that under  $\{H_N\}$  in (5.1)

$$(5.8) \quad \sigma^2(\underline{P}_N) \xrightarrow{P} A^2(1-\rho_{10}-\rho_{01}+\rho_{11}).$$

(5.6), (5.7), (5.8) and the asymptotic normality of  $n^{\frac{1}{2}} \underline{T}_N, jk$  lead to the following theorem.

THEOREM 5.1 If (i)  $\{H_N\}$  in (5.1) holds, (ii) the conditions of theorem 1 of Chernoff and Savage (1958) hold and (iii) the conditions of lemma 7.2 of Puri (1964) hold,  $\underline{\Delta}_N$ , defined by (4.10), has asymptotically a non-central chi-square distribution with  $(p-1)(q-1)$  d.f. and the non-centrality parameter

$$(5.9) \quad \underline{\Delta} = \left[ \frac{1}{pq} \sum_{j=1}^p \sum_{k=1}^q \lambda_{jk}^2 \right] [B^2(H)/A^2(1-\rho_{10}-\rho_{01}+\rho_{11})].$$

Referring back to the model in (2.1), let  $\sigma_u^2$  be the variance of  $U_{ijk}$  and  $\rho_u$  be the correlation between any two  $U_{ijk}$ 's belonging to the same block. Let  $F_{(p-1)(q-1), (n-1)(pq-1)}$  be the classical analysis of variance ratio test statistic for testing  $H_0$  in (2.3) when the parent cdf is assumed to be normal. Then, it can

be shown that under  $\{H_N\}$  in (5.1),  $Q_N = (p-1)(q-1) F_{(p-1)(q-1), (n-1)(pq-1)}$  has asymptotically a non-central chi-square distribution with  $(p-1)(q-1)$  d.f. and the non-centrality parameter

$$(5.10) \quad \Delta_Q = \left[ \frac{1}{pq} \sum_{j=1}^p \sum_{k=1}^q \lambda_{jk}^2 \right] / \sigma_u^2 (1 - \rho_u).$$

Let now  $\sigma_e^2$  be the variance of  $e_{ijk}$ , defined by (2.6). Then, after some simplifications, we obtain from (2.6) that

$$(5.11) \quad \sigma_e^2 = [(p-1)(q-1)/pq] \sigma_u^2 (1 - \rho_u).$$

Consequently, from theorem 5.1, (5.10) and (5.11), we arrive at the following.

THEOREM 5.2 When the conditions of theorem 5.1 hold, the asymptotic relative efficiency (A.R.E.) of the  $\mathcal{S}_N$ -test with respect to the classical analysis of variance test is given by

$$(5.12) \quad e(\{\mathcal{S}_N\}, \{Q_N\}) = \frac{pq}{(p-1)(q-1)(1 - \rho_{10} - \rho_{01} + \rho_{11})} \left[ \sigma_e^2 B^2(H) / A^2 \right].$$

We note that  $\sigma_e^2$  is the variance of the cdf  $H$ , and hence the second factor on the right hand side of (5.12) resembles the usual efficiency factor for the well-known Chernoff-Savage (1958) type of test statistics. Also, it follows from lemmas 4.4 and 4.5 of Sen (1966b) that

$$(5.13) \quad \rho_{10} \geq -1/(p-1), \quad \rho_{01} \geq -1/(q-1),$$

where the equality sign holds iff  $J = H^{-1}$  (apart from an additive constant). Thus, from (5.12) and (5.13), we obtain that

$$(5.14) \quad e(\{\mathfrak{L}_N\}, \{Q_N\}) \geq \frac{1}{\{1 - \frac{1}{pq} [1-(p-1)(q-1)\rho_{11}]\}} \cdot [\sigma_e^2 B^2(H)/A^2] .$$

This leads to the following corollary.

COROLLARY 5.2.1 A sufficient condition for  $e(\{\mathfrak{L}_N\}, \{Q_N\})$  to be at least as large as  $[\sigma_e^2 B^2(H)/A^2]$  is that  $\rho_{11} < 1/(p-1)(q-1)$ .

We shall now consider two special  $\mathfrak{L}_N$ -statistics, namely, the Normal Scores and Wilcoxon scores statistics. In the first case,  $J_N(\frac{\alpha}{N+1})$ , defined by (4.2), is the expected value of the  $\alpha$ -th smallest observation of a sample of size  $N$  from a standard normal distribution, for  $\alpha=1, \dots, N$ . In this case, it is well-known [cf. Chernoff and Savage (1958)] that  $\sigma_e^2 B^2(H)/A^2$  is greater than or equal to 1, where the equality sign holds only when  $H$  is also normal. Thus, the minimum A.R.E. of the normal scores test with respect to the classical analysis of variance test is equal to  $1/\{1 - \frac{1}{pq} [1-(p-1)(q-1)\rho_{11}]\} \geq \frac{1}{2 - (\frac{1}{p} + \frac{1}{q})} > \frac{1}{2}$ . On the other hand, for the class of parent cdf's for which  $\rho_{11} \leq 1/(p-1)(q-1)$ , the normal scores test will be at least as efficient as the  $Q_N$ -test. In particular, if  $H(x)$  is normal,  $\rho_{11} = 1/(p-1)(q-1)$  and  $\sigma_e^2 B^2(H)/A^2 = 1$ , so that the normal scores test and the  $Q_N$ -test become asymptotically power equivalent. For Wilcoxon scores,  $J_N(\frac{\alpha}{N+1}) = \frac{\alpha}{N+1} : 1 \leq \alpha \leq N$ . In this case,  $[\sigma_e^2 B^2(H)/A^2]$  is known to be greater than or equal to 0.864 for all  $H$ . Consequently, the A.R.E. of the Wilcoxon scores test with respect to the  $Q_N$ -test is bounded below by

$$(5.15) \quad 0.864/\{1 - \frac{1}{pq} [1-(p-1)(q-1)\rho_{11}]\} \geq 0.432 .$$

For normal  $F$ , it is known that

$$(5.16) \quad \rho_{10} = \frac{6}{\pi} \sin^{-1} \left( \frac{-1}{2(p-1)} \right), \quad \rho_{01} = \frac{6}{\pi} \sin^{-1} \left( \frac{-1}{2(q-1)} \right), \quad \rho_{11} = \frac{6}{\pi} \sin^{-1} \left( \frac{1}{2(p-1)(q-1)} \right),$$

and hence from (5.12), we obtain that for normal distributions the A.R.E. of the Wilcoxon-scores test with respect to the  $Q_N$ -test is given by

$$(5.17) \quad \frac{3pq}{\pi(p-1)(q-1) \left[ 1 + \frac{6}{\pi} \left\{ \sin^{-1} \frac{1}{2(p-1)} + \sin^{-1} \frac{1}{2(q-1)} + \sin^{-1} \frac{1}{2(p-1)(q-1)} \right\} \right]}$$

The following table illustrates the numerical values of (5.17).

$\begin{matrix} p \\ q \end{matrix}$	2	3	4	5	6	7	8	9	10	15	20	$\infty$
2	.955	.966	.965	.963	.962	.961	.960	.960	.959	.958	.957	.955
3		.975	.974	.972	.971	.961	.970	.970	.969	.968	.968	.966
4			.972	.971	.970	.970	.969	.969	.968	.967	.966	.965
5				.970	.969	.968	.968	.967	.967	.966	.965	.963
6					.969	.967	.966	.966	.966	.964	.964	.962
7						.966	.966	.965	.965	.964	.963	.961
8							.965	.964	.964	.963	.962	.960
9								.964	.964	.962	.962	.960
10									.963	.962	.961	.959
15										.961	.960	.958
20											.959	.957
$\infty$												.955

Thus, the efficiency is bounded below by  $3/\pi$  and may be as high as 0.974.

If we have more than one observation per cell, we may still work with the aligned observations obtained by making adjustments for row, column and grand means. The permutation argument is essentially the same (with  $p$  and  $q$  replaced by  $pr$  and  $qr$ , respectively,  $r$  being the number of observations per cell). For more than two factors, summing over a subset of factors, we may arrive at the desired nuisance parameter-free model and proceed as in this paper. For brevity, the details are omitted.



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