

# NOTE ON THE UTILIZATION OF THE GENERALIZED STUDENT RATIO IN THE ANALYSIS OF VARIANCE OR DISPERSION

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**Summary.** In this note, we shall consider the usage of the generalized Student ratio  $T$  or  $T^2$ , which is the statistic introduced by H. Hotelling [3, 4], in the analysis of variance or dispersion. F. A. Graybill and J. L. Folks [1, 2] have shown that in a randomized block experiment and in the situation of heterogeneity of error variances, the equality of main effect constants can be tested by using the  $T^2$ -statistic. It will be shown that their approach is easily extended for testing the significance of the interaction in a two-factor experiment with different error variances. For the multivariate analysis of dispersion, the use of the generalized  $T^2$  measure is illustrated by a simple example and the asymptotic power of the test based on this statistic is also considered. The tables of 5% and 1% significance points for this test are prepared for three and four dimensional cases.

## 1. Two-factor experiment.

Let us consider that we have a factorial experiment with two factors  $A$  and  $B$ ,  $A$  being at  $p$  levels and  $B$  at  $q$  levels and moreover the pattern composed of those  $pq$  treatments be carried out, for example, at each of  $n$  locations or in each of  $n$  days. We assume that the individual  $x_{i,j,\nu}$  ( $i=1, 2, \dots, p$ ;  $j=1, 2, \dots, q$ ;  $\nu=1, 2, \dots, n$ ) have the mathematical model

$$(1) \quad x_{i,j,\nu} = \mu + \alpha_i + \beta_j + \gamma_{ij} + \delta_\nu + \varepsilon_{i,j,\nu}$$

with ordinary restrictions  $\sum_{i=1}^p \alpha_i = \sum_{j=1}^q \beta_j = \sum_{i=1}^p \gamma_{ij} = \sum_{j=1}^q \gamma_{ij} = 0$ , where  $\mu$  is the general mean,  $\alpha_i$  is the effect of the  $i$  th  $A$ ,  $\beta_j$  is the effect of the  $j$  th  $B$ ,  $\gamma_{ij}$  is the interaction between the  $i$  th  $A$  and the  $j$  th  $B$ ,  $\delta_\nu$  is the effect of the  $\nu$  th location or the  $\nu$  th day (assumed random) and  $\varepsilon_{i,j,\nu}$  ( $\nu=1, 2, \dots, n$ ) are error terms which are assumed to be distributed according to a  $pq$  variate normal distribution with mean  $\mathbf{0}=(0, 0, \dots, 0)$  and with covariance matrix  $\Lambda_1$ . Using the ordinary rule of notation, we shall denote the mean of  $x_{i,j,\nu}$  with respect to some subscripts by  $\bar{x}$  with the subscripts over which the mean operation is made replaced by dots.

Then we can test the hypothesis  $H_1$  that all the  $\alpha_i$  are equal, that is,  $\alpha_1 = \alpha_2 = \dots = \alpha_p (=0)$  in the following way: defining new variables such that

$$(2) \quad \begin{aligned} y_{i,\nu} &= x_{i,\nu} - x_{p,\nu} = \alpha_i - \alpha_p + \varepsilon_{i,\nu} - \varepsilon_{p,\nu} \\ &\quad (i=1, 2, \dots, p-1; \nu=1, 2, \dots, n), \end{aligned}$$

the row vector  $\mathbf{y}_\nu = (y_{1,\nu}, y_{2,\nu}, \dots, y_{(p-1),\nu})$  can be considered a random sample of size  $n$  from a  $(p-1)$ -variate normal population with mean  $(\alpha_1 - \alpha_p, \alpha_2 - \alpha_p, \dots, \alpha_{p-1} - \alpha_p)$  and with covariance matrix  $\Lambda_2$ . Then we can test  $H_1$  by the Hotelling's  $T^2$ -statistic

$$(3) \quad T^2 = (n-1) \bar{\mathbf{y}} \mathbf{S}^{-1} \bar{\mathbf{y}}'$$

where  $\bar{\mathbf{y}} = (y_{1\cdot}, y_{2\cdot}, \dots, y_{(p-1)\cdot})$ ,  $\mathbf{S} = \frac{1}{n} \sum_{\nu=1}^n (\mathbf{y}_\nu - \bar{\mathbf{y}})(\mathbf{y}_\nu - \bar{\mathbf{y}})'$ ,  $\mathbf{S}^{-1}$  is the inverse matrix of  $\mathbf{S}$  and  $\bar{\mathbf{y}}'$  is the transpose of  $\bar{\mathbf{y}}$ , and hence we can also test  $H_1$  by the criterion

$$(4) \quad F = \frac{(n-p+1)}{p-1} \frac{T^2}{n-1} = \frac{(n-p+1)n}{p-1} \bar{\mathbf{y}} \left[ \sum_{\nu=1}^n (\mathbf{y}_\nu - \bar{\mathbf{y}})(\mathbf{y}_\nu - \bar{\mathbf{y}})' \right]^{-1} \bar{\mathbf{y}}'$$

which has Snedecor's  $F$  distribution with  $p-1$  and  $n-p+1$  degrees of freedom under  $H_1$  if  $n > p-1$ .

If we want to test the hypothesis  $H_2$  that  $\beta_1 = \beta_2 = \dots = \beta_q (=0)$ , we can do it in the same way as above.

Next we shall consider the test of the hypothesis  $H_3$  that all the  $\gamma_{ij}$  are zero, the  $\alpha_i$  and the  $\beta_j$  remaining unspecified. To do this we consider the  $(p-1)(q-1)$  new variables such that

$$(5) \quad \begin{aligned} y_{ij\nu} &= x_{ij\nu} - x_{iq\nu} - x_{pj\nu} + x_{pq\nu} \\ &= \gamma_{ij} - \gamma_{iq} - \gamma_{pj} + \gamma_{pq} + (\varepsilon_{ij\nu} - \varepsilon_{iq\nu} - \varepsilon_{pj\nu} + \varepsilon_{pq\nu}) \\ &\quad i=1, 2, \dots, p-1; j=1, 2, \dots, q-1; \nu=1, 2, \dots, n. \end{aligned}$$

It is easily verified that the original hypothesis  $H_3$  is equivalent to the hypothesis  $H'_3$  that  $\phi_{ij} \equiv \gamma_{ij} - \gamma_{iq} - \gamma_{pj} + \gamma_{pq}$  are all zero as follows: if all the  $\gamma_{ij}$  are zero, it is obvious that all the  $\phi_{ij}$  are zero. If  $\phi_{ij} = 0$  for  $i=1, 2, \dots, p-1$  and  $j=1, 2, \dots, q-1$ , we have

$$\sum_{j=1}^{q-1} \phi_{ij} = \sum_{j=1}^{q-1} \gamma_{ij} - (q-1)\gamma_{iq} - \sum_{j=1}^{q-1} \gamma_{pj} + (q-1)\gamma_{pq} = q(\gamma_{pq} - \gamma_{iq}) = 0,$$

since  $\sum_{j=1}^q \gamma_{ij} = 0$ . Therefore  $\gamma_{iq} = \gamma_{pq}$  and  $\gamma_{ij} = \gamma_{pj}$ . Similarly, from  $\sum_{i=1}^{p-1} \phi_{ij} = 0$

and  $\phi_{ij}=0$ , we have  $\gamma_{pj}=\gamma_{pq}$  and  $\gamma_{ij}=\gamma_{iq}$ . Hence  $\gamma_{ij}=\gamma_{iq}=\gamma_{pj}=\gamma_{pq}$ , that is, all the  $\gamma_{ij}$  ( $i=1, 2, \dots, p; j=1, 2, \dots, q$ ) are zero.

Since we can consider the set of  $y_{ij\nu}$  for fixed  $\nu$  as the  $(p-1)(q-1)$ -dimensional normal variate with mean  $\phi_{ij}$  and covariance matrix  $\Lambda_3$ , we can test  $H_3$  by using the Hotelling's  $T^2$

$$(6) \quad T^2=(n-1)\bar{Y} V^{-1} \bar{Y}'$$

and hence by

$$(7) \quad F = \frac{n-(p-1)(q-1)}{(p-1)(q-1)} \frac{T^2}{n-1} \\ = \frac{[n-(p-1)(q-1)]n}{(p-1)(q-1)} \bar{Y} \left[ \sum_{\nu=1}^n (Y_\nu - \bar{Y})(Y_\nu - \bar{Y})' \right]^{-1} \bar{Y}'$$

which has  $F$  distribution with  $(p-1)(q-1)$  and  $\{n-(p-1)(q-1)\}$  degrees of freedom, where  $\bar{Y}$  is the mean of  $Y_\nu$  which is the row vector with components  $y_{ij\nu}$  in some arrangement and  $V$  is the covariance matrix of  $Y_\nu$  ( $\nu=1, 2, \dots, n$ ).

**2. The multivariate analysis of dispersion.**

Let us consider, for example, a randomized block design in which measurements are made in  $p$  dimensions and we assume that individual results  $\mathbf{x}_{ij}=(x_{ij}^{(1)}, x_{ij}^{(2)}, \dots, x_{ij}^{(p)})$  are given by the mathematical model

$$(8) \quad \mathbf{x}_{ij}=\boldsymbol{\mu}+\boldsymbol{\tau}_i+\boldsymbol{\beta}_j+\boldsymbol{\epsilon}_{ij} \quad (i=1, 2, \dots, r; j=1, 2, \dots, s)$$

with  $\sum_{i=1}^r \boldsymbol{\tau}_i = \sum_{j=1}^s \boldsymbol{\beta}_j = \mathbf{0}$ , where

$\boldsymbol{\mu}=(\mu^{(1)}, \mu^{(2)}, \dots, \mu^{(p)})$  is the general mean,  $\boldsymbol{\tau}_i=(\tau_i^{(1)}, \tau_i^{(2)}, \dots, \tau_i^{(p)})$  is the effect of the  $i$  th treatment,  $\boldsymbol{\beta}_j=(\beta_j^{(1)}, \beta_j^{(2)}, \dots, \beta_j^{(p)})$  is the effect of the  $j$  th block, and  $\boldsymbol{\epsilon}_{ij}$ 's are normally and independently distributed around  $\mathbf{0}$  with a covariance matrix  $\Lambda$ . By the usual procedure we have

$$(9) \quad \sum_{i=1}^r \sum_{j=1}^s (\mathbf{x}_{ij}-\mathbf{x}_{..})'(\mathbf{x}_{ij}-\mathbf{x}_{..}) = s \sum_{i=1}^r (\mathbf{x}_{i.}-\mathbf{x}_{..})'(\mathbf{x}_{i.}-\mathbf{x}_{..}) + r \sum_{j=1}^s (\mathbf{x}_{.j}-\mathbf{x}_{..})'(\mathbf{x}_{.j}-\mathbf{x}_{..}) \\ + \sum_{i=1}^r \sum_{j=1}^s (\mathbf{x}_{ij}-\mathbf{x}_{i.}-\mathbf{x}_{.j}+\mathbf{x}_{..})'(\mathbf{x}_{ij}-\mathbf{x}_{i.}-\mathbf{x}_{.j}+\mathbf{x}_{..}) \\ = \mathbf{T} + \mathbf{B} + \mathbf{E}.$$

$\mathbf{E}$  has the Wishart's distribution with  $(r-1)(s-1)$  degrees of freedom and  $\mathbf{E}/(r-1)(s-1)$  is the unbiased estimate of  $\Lambda$ .  $\mathbf{T}$  and  $\mathbf{B}$  have the Wishart's distributions with  $(r-1)$  and  $(s-1)$  degrees of freedom if  $\boldsymbol{\tau}$ 's and  $\boldsymbol{\beta}$ 's are zero, respectively. For testing the hypothesis that the

effects of treatment are all zero, that is,  $H_t: \alpha_1 = \alpha_2 = \dots = \alpha_r (=0)$ , the Wilks' criterion  $W = |E|/|T+E|$ , which is the maximum likelihood criterion, is used in the present practical application. But we can also test this hypothesis based on Hotelling's  $T^2$  as follows: from (9) we have

$$(10) \quad \text{tr } \Lambda^{-1} \left( \sum_{i=1}^r \sum_{j=1}^s (\mathbf{x}_{ij} - \mathbf{x}_{..})' (\mathbf{x}_{ij} - \mathbf{x}_{..}) \right) = \text{tr } \Lambda^{-1} \mathbf{T} + \text{tr } \Lambda^{-1} \mathbf{B} + \text{tr } \Lambda^{-1} \mathbf{E}.$$

It is easily seen that  $\text{tr } \Lambda^{-1} \mathbf{T}$  has  $\chi^2$  distribution with  $(r-1)p$  degrees of freedom when  $H_t$  is true,  $\text{tr } \Lambda^{-1} \mathbf{B}$  has  $\chi^2$ -distribution with  $(s-1)p$  degrees of freedom if all the  $\beta$ 's are zero and  $\text{tr } \Lambda^{-1} \mathbf{E}$  is  $\chi^2$  with  $(r-1)(s-1)p$  degrees of freedom regardless of whether the  $\tau$ 's and  $\beta$ 's are zero or not. Since we have not a priori the knowledge of  $\Lambda$ , we must replace  $\Lambda$  by its unbiased estimate and in our present case we may use  $L = \frac{1}{(r-1)(s-1)} \mathbf{E}$  for it. Then we have the relation

$$(11) \quad \text{tr } L^{-1} \left( \sum_{i=1}^r \sum_{j=1}^s (\mathbf{x}_{ij} - \mathbf{x}_{..})' (\mathbf{x}_{ij} - \mathbf{x}_{..}) \right) = \text{tr } L^{-1} \mathbf{T} + \text{tr } L^{-1} \mathbf{B} + (r-1)(s-1)p.$$

Thus  $\text{tr } L^{-1} \mathbf{T}$  is used as a criterion for testing the hypothesis  $H_t$  and  $\text{tr } L^{-1} \mathbf{B}$  is used as a criterion for testing the hypothesis that  $\beta_1 = \dots = \beta_s (=0)$ . Now we consider the sampling distribution of  $\text{tr } L^{-1} \mathbf{T}$  under the null hypothesis  $H_t$ . It is easily seen that, by suitable orthogonal transformation,  $\text{tr } L^{-1} \mathbf{T}$  may be written as

$$\text{tr } L^{-1} \mathbf{T} = \text{tr} \left( \frac{1}{\alpha} \sum_{\nu=1}^{\alpha} \mathbf{y}'_{\nu} \mathbf{y}_{\nu} \right)^{-1} \left( \sum_{\alpha=1}^{r-1} \mathbf{z}'_{\alpha} \mathbf{z}_{\alpha} \right) = \sum_{\alpha=1}^{r-1} \mathbf{z}'_{\alpha} \left( \frac{1}{\alpha} \sum_{\nu=1}^{\alpha} \mathbf{y}'_{\nu} \mathbf{y}_{\nu} \right)^{-1} \mathbf{z}_{\alpha}$$

where  $\alpha = (r-1)(s-1)$  and  $\mathbf{y}_{\nu} (\nu=1, 2, \dots, r-1)$  and  $\mathbf{z}_{\alpha} (\alpha=1, 2, \dots, a)$  are independently and normally distributed with zero mean vector and with covariance matrix  $\Lambda$ . Therefore  $\text{tr } L^{-1} \mathbf{T}$  may be considered as the sum of  $(r-1)$  Hotelling's  $T^2$  statistics, i.e.,  $T_1^2 + T_2^2 + \dots + T_{r-1}^2$ , components of which are depend on the same unbiased estimate of  $\Lambda$ . The criterions of this type were initially considered by Hotelling [4] and its exact sampling distribution is known when  $p=2$  and moreover, the tables of percentage points of the criterion for testing are now available in this case. For  $p \geq 3$ , the approximate formula of percentage points of the criterion, which was obtained independently by the author [7, 8] and K. Ito [5], is prepared for use. If we denote, in general, the statistic considered above by

$$(12) \quad T_0^2 = m \operatorname{tr} L^{-1}V = \operatorname{tr} L^{-1} \left( \sum_{v=1}^m \mathbf{y}'_v \mathbf{y}_v \right),$$

where  $L$  and  $V$  are two independent unbiased estimates of the population covariance matrix  $\Lambda$  with  $n$  and  $m$  degrees of freedom, respectively. Then the approximate formula of the  $\eta \times 100$  percentage points,  $T_0^2(\eta)$ , is given as

$$(13) \quad \begin{aligned} T_0^2(\eta) = & x^2 + \frac{m}{2n} [p(p+1)(\chi_4 + \chi_2) + mp(\chi_4 - \chi_2)] \\ & + \frac{m}{n^2} \left\{ \frac{m}{16} \left( 1 - \frac{mp-2}{\chi^2} \right) [p(p+1)(\chi_4 + \chi_2) + mp(\chi_4 - \chi_2)]^2 \right. \\ & - \frac{m}{8} [p(p+1)(\chi_4 + \chi_2) + mp(\chi_4 - \chi_2)] [p+1)(x_4 - 1) + mp(\chi_4 - 2\chi_2 + 1)] \\ & - \frac{1}{3} [p(p^2 + 4)(\chi_6 + \chi_4 + \chi_2) + 3mp(p+1)(\chi_6 - \chi_2) + m^2p(\chi_6 - 2\chi_4 + \chi_2)] \\ & + \frac{1}{16} [4p(2p^2 + 5p + 5)(\chi_8 + \chi_6 + \chi_4 + \chi_2) + 16mp(p+1)(\chi_8 - \chi_2) \\ & + mp(p^3 + 2p^2 + 5p + 4)(\chi_8 + \chi_6 - \chi_4 - \chi_2) \\ & \left. + 2m^2p(p^2 + p + 4)(\chi_8 - \chi_6 - \chi_4 + \chi_2) + m^3p^2(\chi_8 - 3\chi_6 + 3\chi_4 - \chi_2)] \right\} \\ & + O(n^{-3}), \end{aligned}$$

where  $\chi^2 \equiv \chi_{mp}^2(\eta)$ , that is,  $\eta \times 100$  percentage points of  $\chi^2$  distribution with  $mp$  degrees of freedom, and  $\chi_{2s} \equiv \chi_{mp}^{2s}(\eta) / mp(mp+2) \cdots (mp+2s-2)$ . Table I and Table II give the values of  $T_0^2(\eta)$  calculated by the above formula for  $p=3, 4$  and  $\eta=0.01, 0.05$ .

2.1. *The power of the test based on the statistic  $T_0^2$ .* In order to evaluate the power function of the test based on the statistic  $\operatorname{tr} L^{-1}T$  or  $\operatorname{tr} L^{-1}B$ , we shall consider the sampling distribution of the general criterion  $T_0^2 = m \operatorname{tr} L^{-1}V = \operatorname{tr} L^{-1} \left( \sum_{v=1}^m \mathbf{y}'_v \mathbf{y}_v \right) = \sum_{v=1}^m \mathbf{y}_v L^{-1} \mathbf{y}'_v$  under the hypothesis that  $\mathbf{y}_v$  has the mean vector  $\boldsymbol{\eta}_v$ , all of which are not zero. If we have so large value of  $n$  that  $L$  may be replaced by the population covariance matrix  $\Lambda$ , we can consider  $\chi_0^2 = \operatorname{tr} \Lambda^{-1} \left( \sum_{v=1}^m \mathbf{y}'_v \mathbf{y}_v \right)$  instead of  $T_0^2$ . In this case,  $\chi_0^2$  is distributed according to the non-central chi-square distribution with  $mp$  degrees of freedom and with the parameter  $\delta$ , where

$$(14) \quad \delta^2 = \sum_{v=1}^m \boldsymbol{\eta}_v \Lambda^{-1} \boldsymbol{\eta}'_v,$$

since  $\chi_0^2 = \sum_{\nu=1}^m \chi_{0\nu}^2 = \sum_{\nu=1}^m \mathbf{y}_\nu \Lambda^{-1} \mathbf{y}'_\nu$ ,  $\chi_{0\nu}^2$  has the non-central chi-square distribution with  $p$  degrees of freedom and with the parameter  $\delta_\nu = \{\boldsymbol{\eta}_\nu \Lambda^{-1} \boldsymbol{\eta}'_\nu\}^{\frac{1}{2}}$  and  $\chi_{0\nu}^2 (\nu=1, 2, \dots, m)$  are mutually independent.

When  $n$  is not so large, the more accurate formula is necessary, which will be considered from now on. From the distribution of  $\chi_0^2$ , we have

$$(15) \quad P_r \{m \operatorname{tr} \Lambda^{-1} \mathbf{V} \leq 2\xi\} = G_p(\xi; \delta),$$

TABLE Ia  $T_0^2(0.05)$  for  $p=3$

$m \backslash n$	10	12	14	16	18	20	22	24	26	28	30
1	15.25	13.35	12.22	11.46	10.93	10.53	10.22	9.98	9.78	9.61	9.47
2	23.32	21.04	19.53	18.47	17.68	17.07	16.59	16.21	15.88	15.61	15.39
3	31.94	28.73	26.61	25.12	24.01	23.16	22.49	21.95	21.50	21.12	20.80
4	40.46	36.28	33.52	31.58	30.14	29.04	28.17	27.47	26.89	26.41	25.99
5	48.44	43.40	40.08	37.74	36.01	34.68	33.63	32.78	32.08	31.50	31.00
6	56.52	50.56	46.64	43.88	41.84	40.27	39.03	38.03	37.21	36.52	35.93
7	64.53	57.66	53.14	49.95	47.60	45.79	44.37	43.22	42.27	41.47	40.79
8	72.50	64.72	59.59	55.98	53.31	51.27	49.65	48.35	47.27	46.37	45.61
9	80.43	71.73	66.00	61.97	58.99	56.70	54.90	53.44	52.24	51.23	50.38
10	88.33	78.72	72.38	67.93	64.63	62.11	60.12	58.51	57.18	55.96	55.12
12	104.0	92.60	85.07	79.77	75.85	72.85	70.47	68.56	66.98	65.65	64.53
14	119.7	106.4	97.69	91.54	87.00	83.51	80.76	78.53	76.70	75.16	73.85
16	135.1	120.1	110.2	103.2	98.08	94.12	90.99	88.46	86.37	84.62	83.13
18	150.5	133.8	122.7	114.9	109.1	104.7	101.2	98.34	95.99	94.03	91.36
20	165.8	147.3	135.1	126.5	120.1	115.2	111.3	108.2	105.6	103.4	101.6

$m \backslash n$	35	40	45	50	60	80	100
1	9.18	8.99	8.85	8.74	8.57	8.37	8.26
2	14.94	14.61	14.37	14.17	13.89	13.55	13.35
3	20.18	19.73	19.48	19.11	18.72	18.24	17.97
4	25.19	24.61	24.17	23.82	23.32	22.71	22.36
5	30.03	29.33	28.80	28.38	27.77	27.03	26.61
6	34.79	33.96	33.33	32.85	32.13	31.27	30.76
7	39.48	38.53	37.81	37.25	36.42	35.43	34.85
8	44.12	43.04	42.23	41.59	40.66	39.53	38.88
9	48.72	47.52	46.60	45.89	44.85	43.59	42.86
10	53.28	51.96	50.95	50.16	49.01	47.62	46.81
12	62.34	60.76	59.56	58.62	57.25	55.59	54.63
14	71.31	69.47	68.08	66.99	65.39	63.46	62.34
16	80.23	78.14	76.54	75.30	73.48	71.27	69.99
18	89.11	86.75	84.97	83.57	81.52	79.04	77.59
20	97.95	95.34	93.35	91.80	89.52	86.76	85.15

where  $\rho \equiv mp/2$  and

$$(16) \quad G_\rho(\xi; \delta) = e^{-\frac{\xi}{2}} \sum_{j=0}^{\infty} \frac{(\delta^2/2)^j}{j! \Gamma(\rho+j)} \int_0^\xi t^{\rho+j-1} e^{-t} dt .$$

According to the Welch-James' method [6, 9] which was used to obtain the formula (13), we try to evaluate

$$P_r\{m \operatorname{tr} L^{-1} V \leq 2\xi\} = \left[ 1 + \frac{1}{n} \sum_{rstu} \lambda_{ur} \lambda_{st} \delta_{rs} \delta_{tu} \right]$$

TABLE Ib  $T_0^2(0.01)$  for  $p=3$

$m \backslash n$	10	12	14	16	18	20	22	24	26	28	30
1	28.47	23.59	20.83	19.08	17.86	16.97	16.30	15.76	15.33	14.98	14.68
2	36.80	32.40	29.52	27.50	26.01	24.88	23.99	23.27	22.68	22.18	21.76
3	48.24	42.35	38.49	35.80	33.82	32.32	31.13	30.17	29.39	28.73	28.18
4	59.19	51.85	47.05	43.70	41.24	39.37	37.90	36.72	36.24	35.43	34.74
5	69.88	61.10	55.37	51.37	48.44	46.21	44.46	43.05	41.89	40.92	40.11
6	80.41	70.20	63.55	58.90	55.50	52.91	50.87	49.24	47.90	46.78	45.83
7	90.83	79.20	71.62	66.33	62.46	59.51	57.19	55.33	53.81	52.53	51.46
8	101.2	88.11	79.61	73.69	69.34	66.04	63.44	61.35	59.64	58.22	57.01
9	111.5	96.98	87.56	80.99	76.17	72.51	69.63	67.32	65.42	63.84	62.50
10	121.7	105.8	95.45	88.24	82.96	78.94	75.78	73.24	71.16	69.42	67.95
12	142.1	123.4	111.2	102.7	96.49	91.74	88.02	85.03	82.57	80.52	78.79
14	162.2	140.8	126.8	117.0	109.8	104.4	100.1	96.65	93.82	91.47	89.47
16	182.2	158.1	142.2	131.1	123.1	116.9	112.1	108.2	105.0	102.3	100.1
18	202.0	175.1	157.6	145.3	136.3	129.4	124.0	119.7	116.1	113.1	110.6
20	221.5	192.0	172.7	159.2	149.3	141.8	135.8	131.1	127.1	123.8	121.1

$m \backslash n$	35	40	45	50	60	80	100
1	14.05	13.67	13.39	13.16	12.83	12.44	12.21
2	20.95	20.37	19.93	19.58	18.84	18.24	17.89
3	27.11	26.33	25.75	25.29	24.63	23.83	23.37
4	32.92	31.96	31.24	30.68	29.86	28.88	28.31
5	38.53	37.39	36.53	35.87	34.90	33.73	33.06
6	44.00	42.68	41.69	40.92	39.79	38.45	37.67
7	49.37	47.88	46.75	45.87	44.60	43.07	42.19
8	54.68	53.00	51.74	50.75	49.32	47.61	46.63
9	59.92	58.06	56.66	55.57	53.99	52.09	51.00
10	65.12	63.08	61.54	60.35	58.61	56.53	55.33
12	75.44	73.03	71.22	69.81	67.75	65.30	63.88
14	85.61	82.84	80.75	79.12	76.75	73.91	72.28
16	95.71	92.56	90.19	88.35	85.66	82.44	80.58
18	105.7	102.2	99.57	97.51	94.50	90.89	88.81
20	115.7	111.8	108.9	106.6	103.3	99.26	96.95

$$(17) \quad + \frac{1}{n^2} \left\{ \frac{4}{3} \sum_{rstuvw} \lambda_{wr} \lambda_{st} \lambda_{uo} \partial_{rs} \partial_{tu} \partial_{vw} + \frac{1}{2} \sum_{rstuvwxy} \lambda_{ur} \lambda_{st} \lambda_{yv} \lambda_{wx} \partial_{rs} \partial_{tu} \partial_{vw} \partial_{xy} \right\} + O(n^{-3}) \Big] P_r \{ m \operatorname{tr} \Lambda^{-1} V \leq 2\xi \} ,$$

where  $\partial_{rs} = \frac{1}{2} (1 + \delta_{rs}) \frac{\partial}{\partial \lambda_{rs}}$ , ( $\delta_{rs}$  is Kronecker's  $\delta$ ), and summations are taken over  $r, s, \dots = 1, 2, \dots p$ .

2.2. *The evaluation of the derivatives of  $P_r \{ m \operatorname{tr} \Lambda^{-1} V \leq 2\xi \}$ .* We

**TABLE IIa**  $T_0^2(0.05)$  for  $p=4$

$m \backslash n$	10	12	14	16	18	20	22	24	26	28	30
1	23.54	19.38	17.09	15.65	14.67	13.95	13.41	12.98	12.64	12.36	12.12
2	32.85	29.04	26.55	24.80	23.51	22.53	21.75	21.13	20.62	20.19	19.82
3	45.11	39.80	36.32	33.89	32.10	30.73	29.65	28.79	28.08	27.48	26.98
4	56.96	50.17	45.74	42.63	40.35	39.61	37.24	36.14	35.24	34.48	33.84
5	68.59	60.34	54.95	51.19	48.42	46.31	44.65	43.32	42.22	41.30	40.52
6	80.09	70.39	64.05	59.62	56.37	53.89	51.94	50.37	49.08	48.01	47.10
7	91.51	80.35	73.07	67.98	64.24	61.39	59.15	57.35	55.87	54.63	53.59
8	102.94	90.32	82.07	76.32	72.09	68.87	66.34	64.30	62.63	61.23	60.05
9	114.4	100.3	91.08	84.64	79.92	76.33	73.50	71.22	69.36	67.80	66.48
10	125.3	109.8	99.72	92.67	87.49	83.54	80.43	77.94	75.89	74.17	72.72
12	147.9	129.5	117.5	109.1	102.9	98.20	94.51	91.53	89.09	87.06	85.33
14	170.3	148.9	135.0	125.2	118.2	112.7	108.5	105.0	102.2	99.84	97.85
16	192.5	168.3	152.5	141.5	133.4	127.2	122.4	118.5	115.2	113.6	110.3
18	214.6	187.6	170.0	157.6	148.6	141.6	136.2	131.8	128.2	125.2	122.7
20	236.6	206.8	187.3	173.7	163.7	156.0	150.0	145.2	141.2	137.9	135.0

$m \backslash n$	35	40	45	50	60	80	100
1	11.62	11.32	11.09	10.92	10.66	10.35	10.16
2	19.12	18.61	18.22	17.92	17.48	16.95	16.65
3	26.00	25.29	24.76	24.35	23.74	23.01	22.59
4	32.60	31.70	31.03	30.50	29.73	28.81	28.28
5	39.02	37.94	37.12	36.49	35.56	34.44	33.80
6	45.33	44.06	43.11	42.36	41.27	39.97	39.21
7	51.56	50.11	49.01	48.15	46.90	45.41	44.54
8	57.76	56.11	54.87	53.90	52.50	50.81	49.83
9	63.92	62.08	60.70	59.62	58.05	56.16	55.07
10	69.92	67.90	66.38	65.19	63.47	61.39	60.20
12	82.00	79.60	77.79	76.38	74.32	71.86	70.44
14	93.98	91.20	89.10	87.47	85.09	82.23	80.58
16	105.89	102.73	100.34	98.48	95.77	92.52	90.64
18	117.8	114.2	111.5	109.4	106.4	102.8	100.6
20	129.6	125.6	122.7	120.4	117.0	112.9	110.6



consider

$$(18) \quad J = P_r \{ m \operatorname{tr} (\Lambda + \epsilon)^{-1} V \leq 2\xi \} ,$$

where  $\epsilon$  is a symmetric matrix composing of the small increments  $\epsilon_{ij}$  to  $\lambda_{ij}$ . Then, by Taylor's theorem.

$$(19) \quad J = \left[ 1 + \sum_{rs} \epsilon_{rs} \partial_{rs} + \frac{1}{2} \sum_{rstu} \epsilon_{rs} \epsilon_{tu} \partial_{rs} \partial_{tu} + \dots \right] P_r \{ m \operatorname{tr} \Lambda^{-1} V \leq 2\xi \} .$$

On the other hand we can also express  $J$  in the form

**TABLE IIb**  $T_0^2(0.01)$  for  $p=4$

$m \backslash n$	10	12	14	16	18	20	22	24	26	28	30
1	44.84	34.25	28.86	25.63	23.49	21.97	20.84	19.97	19.28	18.71	18.25
2	50.26	43.42	48.98	35.89	33.63	31.92	30.57	29.49	28.61	27.87	27.25
3	66.01	56.94	51.06	46.98	43.99	41.73	39.96	38.54	37.37	36.40	35.59
4	81.09	69.86	62.59	57.55	53.87	51.07	48.89	47.13	45.70	44.51	43.50
5	95.81	82.47	73.83	67.84	63.46	60.15	57.55	55.47	53.77	52.36	51.17
6	110.33	94.88	84.88	77.94	72.88	69.05	66.05	63.65	61.68	60.05	58.67
7	124.70	107.15	95.80	87.92	82.18	77.83	74.43	72.70	69.47	67.62	65.06
8	139.17	119.49	106.76	97.93	91.50	86.62	82.81	79.75	77.26	75.18	73.43
9	153.44	131.65	117.56	107.79	100.67	95.28	91.06	87.68	84.91	82.62	80.68
10	167.31	143.50	128.09	117.41	109.63	103.73	99.11	95.42	92.40	89.88	87.76
12	195.55	167.56	149.45	136.89	127.74	120.80	115.37	111.03	107.38	104.52	102.03
14	223.51	191.39	170.60	156.19	145.67	137.70	147.47	126.47	122.39	118.99	116.13
16	251.29	215.08	191.63	175.37	163.50	154.50	147.46	141.82	137.21	133.37	130.13
18	278.89	238.63	212.55	194.45	181.24	171.22	163.38	157.09	151.95	147.67	144.06
20	306.18	261.96	233.29	213.39	198.85	187.82	179.18	172.26	166.59	161.87	157.89

$m \backslash n$	35	40	45	50	60	80	100
1	17.22	16.65	16.23	15.92	15.44	14.85	14.51
2	26.05	25.19	24.54	24.03	23.31	22.44	21.93
3	34.01	32.88	32.03	31.37	30.41	29.27	28.62
4	41.56	40.17	39.13	38.32	37.14	35.74	34.94
5	48.87	47.22	45.99	45.02	43.64	41.98	41.04
6	56.01	54.11	52.69	51.58	49.98	48.07	46.98
7	63.04	60.89	59.27	58.02	56.21	54.04	52.80
8	70.05	67.64	65.53	64.13	62.09	59.67	58.29
9	76.95	74.28	72.28	70.72	68.47	65.80	64.26
10	83.68	80.76	78.58	76.88	74.42	71.49	69.81
12	97.23	93.79	91.22	89.22	86.33	82.88	80.91
14	110.6	106.7	103.7	101.4	98.07	94.12	91.83
16	123.9	119.4	116.1	113.5	109.7	105.2	102.6
18	137.1	132.1	128.4	125.5	121.3	116.8	113.4
20	150.2	144.7	140.6	137.4	133.7	127.2	124.0

$$(20) \quad J = (2\pi)^{-p} |\Lambda|^{-\frac{m}{2}} \int_R \exp \left\{ -\frac{1}{2} \sum_{\alpha=1}^m (\mathbf{y}_\alpha - \boldsymbol{\eta}_\alpha) \Lambda^{-1} (\mathbf{y}_\alpha - \boldsymbol{\eta}_\alpha)' \right\} \prod_{\alpha} d\mathbf{y}_\alpha,$$

where  $R$  is the domain defined by

$$(21) \quad R : \sum_{\alpha=1}^m \mathbf{y}_\alpha (\Lambda + \varepsilon)^{-1} \mathbf{y}'_{\alpha} \leq 2\xi.$$

Now we make the non-singular linear transformations

$$\mathbf{y}_\alpha = \mathbf{Z}_\alpha \mathbf{C}, \quad \boldsymbol{\eta}_\alpha = \boldsymbol{\zeta}_\alpha \mathbf{C}, \quad \alpha = 1, 2, \dots, m$$

such that

$$\frac{1}{2} \mathbf{C} (\Lambda + \varepsilon)^{-1} \mathbf{C}' = \mathbf{I}, \quad \frac{1}{2} \mathbf{C} \Lambda^{-1} \mathbf{C}' = \mathbf{I} - \boldsymbol{\gamma},$$

where  $\boldsymbol{\gamma}$  is a diagonal matrix,  $\text{diag} \{ \gamma_1, \gamma_2, \dots, \gamma_p \}$ . Under these transformations  $J$  becomes

$$(22) \quad J = \pi^{-p} |\mathbf{I} - \boldsymbol{\gamma}|^{-\frac{m}{2}} \int_{R'} \exp \left\{ -\sum_{\alpha=1}^m (\mathbf{Z}_\alpha - \boldsymbol{\zeta}_\alpha) (\mathbf{I} - \boldsymbol{\gamma}) (\mathbf{Z}_\alpha - \boldsymbol{\zeta}_\alpha)' \right\} \prod_{\alpha} d\mathbf{Z}_\alpha$$

$$R' : \sum_{\alpha=1}^m \mathbf{Z}_\alpha \mathbf{Z}'_{\alpha} \leq \xi.$$

Furthermore this integration can be expressed in the very simple form as below :

$$J = \pi^{-p} |\mathbf{I} - \boldsymbol{\gamma}|^{\frac{m}{2}} \int_{R'} \exp \left\{ -\sum_{\alpha=1}^m \mathbf{Z}_\alpha \mathbf{Z}'_{\alpha} + \sum_{\alpha=1}^m \mathbf{Z}_\alpha \boldsymbol{\gamma}' \mathbf{Z}'_{\alpha} + 2 \sum_{\alpha=1}^m \mathbf{Z}_\alpha (\mathbf{I} - \boldsymbol{\gamma}) \boldsymbol{\zeta}'_{\alpha} - \sum_{\alpha=0}^m \boldsymbol{\zeta}'_{\alpha} (\mathbf{I} - \boldsymbol{\gamma}) \boldsymbol{\zeta}_{\alpha} \right\} \prod_{\alpha} d\mathbf{y}_\alpha$$

$$= \pi^{-p} |\mathbf{I} - \boldsymbol{\gamma}|^{\frac{m}{2}} e^{-\frac{\delta^2}{2}} \sum_{\nu_{11}, \dots, \nu_{mp}=0}^{\infty} \sum_{\mu_{11}, \dots, \mu_{mp}=0}^{\infty} \frac{\prod_{\alpha=1}^m \prod_{i=1}^p \gamma_i^{\nu_{\alpha i}} (1 - \gamma_i)^{\mu_{\alpha i}} 2^{\mu_{\alpha i}} \xi^{\mu_{\alpha i}}}{\prod_{\alpha=1}^m \prod_{i=1}^p \nu_{\alpha i}! \mu_{\alpha i}!}$$

$$\times \int_{R'} \prod_{\alpha=1}^m \prod_{i=1}^p Z_{\alpha i}^{2\nu_{\alpha i} + \mu_{\alpha i}} e^{-\sum_{\alpha=1}^m \sum_{i=1}^p Z_{\alpha i}^2} \prod_{\alpha=1}^m \prod_{i=1}^p dZ_{\alpha i}$$

$$= \pi^{-p} |\mathbf{I} - \boldsymbol{\gamma}|^{\frac{m}{2}} e^{-\frac{\delta^2}{2}} \sum_{\nu_{11}, \dots, \nu_{mp}=0}^{\infty} \sum_{\mu_{11}, \dots, \mu_{mp}=0}^{\infty} \prod_{\alpha=1}^m \prod_{i=1}^p \frac{\gamma_i^{\nu_{\alpha i}} (1 - \gamma_i)^{2\mu_{\alpha i}} 2^{2\mu_{\alpha i}} \xi^{2\mu_{\alpha i}}}{\nu_{\alpha i}! (2\mu_{\alpha i})!}$$

$$\times \int \sum_{\alpha=1}^m \sum_{i=1}^p t_{\alpha i} \prod_{\alpha=1}^m \prod_{i=1}^p t_{\alpha i}^{\nu_{\alpha i} + \mu_{\alpha i} - \frac{1}{2}} e^{-\sum_{\alpha=1}^m \sum_{i=1}^p t_{\alpha i}} \prod_{\alpha, i} dt_{\alpha i}$$

$$= |\mathbf{I} - \boldsymbol{\gamma}|^{\frac{m}{2}} e^{-\frac{\delta^2}{2}} \prod_{\alpha=1}^m \prod_{i=1}^p \sum_{\nu_{\alpha i}, \mu_{\alpha i}=0}^{\infty} \frac{\gamma_i^{\nu_{\alpha i}} (1 - \gamma_i)^{2\mu_{\alpha i}} \xi^{2\mu_{\alpha i}}}{\nu_{\alpha i}! \mu_{\alpha i}!} \frac{\Gamma\left(\nu_{\alpha i} + \mu_{\alpha i} + \frac{1}{2}\right)}{\Gamma\left(\mu_{\alpha i} + \frac{1}{2}\right)}$$

$$\begin{aligned} & \times \frac{1}{\Gamma\left[\sum_{\alpha,i}(\nu_{\alpha i} + \mu_{\alpha i}) + \rho\right]} \int_0^\xi t^{\sum_{\alpha,i}(\nu_{\alpha i} + \mu_{\alpha i}) + \rho - 1} e^{-t} dt \\ & = |\mathbf{I} - \boldsymbol{\gamma}|^{\frac{m}{2}} e^{-\frac{\delta^2}{2}} \prod_{\alpha=1}^m \prod_{i=1}^p \sum_{\mu_{\alpha i}=0}^{\infty} \frac{(1-\gamma_i)^{2\mu_{\alpha i}} \xi^{2\mu_{\alpha i}}}{\mu_{\alpha i}!} \\ & \quad \times \sum_{\nu_{\alpha i}=0}^{\infty} \frac{\gamma_i^{\nu_{\alpha i}}}{\nu_{\alpha i}!} \left(\nu_{\alpha i} + \mu_{\alpha i} - \frac{1}{2}\right) \left(\nu_{\alpha i} + \mu_{\alpha i} - \frac{3}{2}\right) \cdots \left(\mu_{\alpha i} + \frac{1}{2}\right) G_{\sum(\nu_{\alpha i} + \mu_{\alpha i}) + \rho}(\xi), \end{aligned}$$

where  $G_\alpha(\xi) = [\Gamma(\alpha)]^{-1} \int_0^\xi t^{\alpha-1} e^{-t} dt$ . If we define an operator  $E$  for fixed  $\xi$  and for any positive integer  $f$  such that  $E^f G_\rho(\xi) = G_{\rho+f}(\xi)$ , we can write

$$\begin{aligned} & \prod_{\alpha=1}^m \prod_{i=1}^p \sum_{\nu_{\alpha i}=0}^{\infty} \frac{\gamma_i^{\nu_{\alpha i}}}{\nu_{\alpha i}!} \left(\nu_{\alpha i} + \mu_{\alpha i} - \frac{1}{2}\right) \left(\nu_{\alpha i} + \mu_{\alpha i} - \frac{3}{2}\right) \cdots \left(\mu_{\alpha i} + \frac{1}{2}\right) G_{\sum(\nu_{\alpha i} + \mu_{\alpha i}) + \rho}(\xi) \\ & = \prod_{\alpha=1}^m \prod_{i=1}^p (1 - \gamma_i E)^{-\left(\mu_{\alpha i} + \frac{1}{2}\right)} G_{\sum \mu_{\alpha i} + \rho}(\xi). \end{aligned}$$

Therefore

$$\begin{aligned} (27) \quad J & = \left\{ \frac{|\mathbf{I} - \boldsymbol{\gamma} E|}{|\mathbf{I} - \boldsymbol{\gamma}|} \right\}^{-\frac{m}{2}} e^{-\frac{\delta^2}{2}} \sum_{\mu_{11}, \dots, \mu_{mp}} \frac{\prod_{\alpha=1}^m \prod_{i=1}^p (1 - \gamma_i)^{2\mu_{\alpha i}} \zeta_{\alpha i}^{2\mu_{\alpha i}} (1 - \gamma_i E)^{-\mu_{\alpha i}}}{\prod_{\alpha=1}^m \prod_{i=1}^p \mu_{\alpha i}!} G_{\sum \mu_{\alpha i} + \rho}(\xi) \\ & = \left\{ \frac{|\mathbf{I} - \boldsymbol{\gamma} E|}{|\mathbf{I} - \boldsymbol{\gamma}|} \right\}^{-\frac{m}{2}} e^{-\frac{\delta^2}{2}} \sum_{j=0}^{\infty} \frac{1}{j!} \left[ \sum_{\alpha=1}^m \sum_{i=1}^p (1 - \gamma_i)^2 \zeta_{\alpha i}^2 (1 - \gamma_i E)^{-1} \right]^j G_{\rho+j}(\xi) \\ & = \left\{ \frac{|\mathbf{I} - \boldsymbol{\gamma} E|}{|\mathbf{I} - \boldsymbol{\gamma}|} \right\}^{-\frac{m}{2}} e^{-\frac{\delta^2}{2}} \sum_{j=0}^{\infty} \frac{1}{j!} \left[ \sum_{\alpha=1}^m \zeta_\alpha (\mathbf{I} - \boldsymbol{\gamma}) (\mathbf{I} - \boldsymbol{\gamma} E)^{-1} (\mathbf{I} - \boldsymbol{\gamma}) \zeta'_\alpha \right]^j G_{\rho+j}(\xi). \end{aligned}$$

Since

$$(24) \quad \frac{|\mathbf{I} - \boldsymbol{\gamma} E|}{|\mathbf{I} - \boldsymbol{\gamma}|} = \frac{|\Lambda^{-1} - [(\Lambda + \boldsymbol{\epsilon})^{-1} - \Lambda^{-1}] \Delta|}{|\Lambda^{-1}|} = |\mathbf{I} - [(\Lambda + \boldsymbol{\epsilon})^{-1} \Lambda - \mathbf{I}] \Delta|$$

and

$$(25) \quad \zeta_\alpha (\mathbf{I} - \boldsymbol{\gamma}) (\mathbf{I} - \boldsymbol{\gamma} E)^{-1} (\mathbf{I} - \boldsymbol{\gamma}) \zeta'_\alpha = \frac{1}{2} \eta_\alpha [\mathbf{I} - \{(\Lambda + \boldsymbol{\epsilon})^{-1} \Lambda - \mathbf{I}\} \Delta]^{-1} \Lambda^{-1} \eta'_\alpha,$$

we have the expression

$$(26) \quad J = \{|\mathbf{I} - \mathbf{X} \Delta|\}^{-\frac{m}{2}} e^{-\frac{\delta^2}{2}} \sum_{j=0}^{\infty} \frac{1}{j!} \left[ \frac{1}{2} \sum_{\alpha=1}^m \eta_\alpha (\mathbf{I} - \mathbf{X} \Delta)^{-1} \Lambda^{-1} \eta'_\alpha \right]^j G_{\rho+j}(\xi),$$

where  $\Delta \equiv E - 1$  and  $\mathbf{X} = (\Lambda + \boldsymbol{\epsilon})^{-1} \Lambda - \mathbf{I}$ .

In order to obtain the derivatives of  $P_r \{m \operatorname{tr} \Lambda^{-1} \mathbf{V} \leq 2\xi\}$ , we must carry out the expansion of the above  $J$  in powers of  $\epsilon_{rs}$  and compare

the result of the expansion with (19). Since

$$\{|I - X\Delta|\}^{-\frac{m}{2}} = 1 + \frac{m}{2} \operatorname{tr} X\Delta + \frac{m}{4} \left[ \operatorname{tr}(X\Delta)^2 + \frac{m}{2} (\operatorname{tr} X\Delta)^2 \right] + \dots,$$

and  $(I - X\Delta)^{-1} = I + X\Delta + (X\Delta)^2 + \dots,$

we have, noting  $\delta^2 = \sum_{\alpha=1}^m \eta_\alpha \Lambda^{-1} \eta'_\alpha$

$$\begin{aligned} J &= \left\{ 1 + \frac{m}{2} \operatorname{tr} X\Delta + \frac{m}{4} \left[ \operatorname{tr}(X\Delta)^2 + \frac{m}{2} (\operatorname{tr} X\Delta)^2 \right] + \frac{m}{4} (\operatorname{tr} X\Delta) \sum_{\alpha=1}^m \eta_\alpha X \Lambda^{-1} \eta'_\alpha (\Delta^2 + \Delta) \right. \\ (27) \quad &+ \frac{1}{2} \sum_{\alpha=1}^m \eta_\alpha X \Lambda^{-1} \eta'_\alpha (\Delta^2 + \Delta) + \frac{1}{2} \sum_{\alpha=1}^m \eta_\alpha X^2 \Lambda^{-1} \eta'_\alpha (\Delta^3 + \Delta^2) \\ &+ \frac{1}{8} \left( \sum_{\alpha=1}^m \eta_\alpha X \Lambda^{-1} \eta'_\alpha \right)^2 (\Delta^4 + 2\Delta^3 + \Delta^2) \\ &\left. + (\text{terms of higher powers than } X^2) \right\} G_p(\xi; \delta). \end{aligned}$$

Now we interpret  $E$  as the operator such that  $E^j G_p(\xi; \delta) = G_{p+j}(\xi; \delta)$ .

Let us use the abbreviated notations

$$\begin{aligned} \Lambda_{rs} &= \partial_{rs} \Lambda = \frac{1}{2} (1 + \delta_{rs}) \frac{\partial}{\partial \lambda_{rs}} \Lambda, \quad [rs] = \operatorname{tr} \Lambda^{-1} \Lambda_{rs}, \quad [rs|tu] = \operatorname{tr} \Lambda^{-1} \Lambda_{rs} \Lambda^{-1} \Lambda_{tu}, \dots \\ \delta_{(rs)}^2 &= \partial_{rs} \delta^2 = \partial_{rs} \sum_{\alpha=1}^m \eta_\alpha \Lambda^{-1} \eta'_\alpha = \sum_{\alpha=1}^m \eta_\alpha \Lambda^{-1} \Lambda_{rs} \Lambda^{-1} \eta'_\alpha, \\ \delta_{(rs,tu)}^2 &= \sum_{\alpha=1}^m \eta_\alpha \Lambda^{-1} \Lambda_{rs} \Lambda^{-1} \Lambda_{tu} \Lambda^{-1} \eta'_\alpha, \dots \end{aligned}$$

Since  $X$  has the expansion

$$X = - \sum_{rs} \epsilon_{rs} \Lambda^{-1} \Lambda_{rs} + \sum_{rstu} \epsilon_{rs} \epsilon_{tu} \Lambda^{-1} \Lambda_{rs} \Lambda^{-1} \Lambda_{tu} - \sum_{rstuvw} \epsilon_{rs} \epsilon_{tu} \epsilon_{vw} \Lambda^{-1} \Lambda_{rs} \Lambda^{-1} \Lambda_{tu} \Lambda^{-1} \Lambda_{vw} + \dots$$

$J$  can be written

$$(28) \quad J = \{ 1 - \sum \epsilon_{rs} (F_1) + \sum \epsilon_{rs} \epsilon_{tu} (F_2) - \dots \} G_p(\xi; \delta),$$

where

$$\begin{aligned} (F_1) &= \frac{m}{2} [rs] \Delta + \frac{1}{2} \delta_{(rs)}^2 (\Delta^2 + \Delta), \\ (F_2) &= \frac{m}{2} [rs|tu] \left( \Delta + \frac{1}{2} \Delta^2 \right) + \frac{m^2}{8} [rs][tu] \Delta^2 + \frac{m}{4} [rs] \delta_{(tu)}^2 (\Delta^3 + \Delta^2) \\ &\quad + \frac{1}{2} \delta_{(rs,tu)}^2 (\Delta^3 + 2\Delta^2 + \Delta) + \frac{1}{8} \delta_{(rs)}^2 \delta_{(tu)}^2 (\Delta^4 + 2\Delta^3 + \Delta^2) \end{aligned}$$

and so on. Thus we have

$$\begin{aligned}
 \partial_{rs} P_r \{m \operatorname{tr} \Lambda^{-1} \mathbf{V} \leq 2\xi\} &= - \left\{ \frac{m}{2} [rs] \Delta + \frac{1}{2} \delta_{(rs)}^2 (\Delta^2 + \Delta) \right\} G_\rho(\xi; \delta) \\
 (29) \qquad \qquad \qquad &= \left\{ \frac{m}{2} [rs] + \frac{1}{2} \delta_{(rs)}^2 E^2 \right\} g_\rho(\xi; \delta),
 \end{aligned}$$

$$\begin{aligned}
 \partial_{rs} \partial_{tu} P_r \{m \operatorname{tr} \Lambda^{-1} \mathbf{V} \leq 2\xi\} &= - \left\{ \frac{m}{2} [rs|tu](E^2 + E) + \frac{m^2}{4} [rs][tu](E^2 - E) \right. \\
 (30) \qquad \qquad \qquad &+ \frac{m}{2} [rs] \delta_{(tu)}^2 (E^3 - E^2) + \delta_{(rs, tu)}^2 E^3 \\
 &+ \left. \frac{1}{4} \delta_{(rs)}^2 \delta_{(tu)}^2 (E^4 - E^3) \right\} g_\rho(\xi; \delta),
 \end{aligned}$$

and so on, where

$$\begin{aligned}
 g_\rho(\xi; \delta) &= D_\rho(\xi; \delta) = e^{-\frac{\delta^2}{2}} \sum_{j=0}^{\infty} \frac{\left(\frac{\delta^2}{2}\right)^j}{j!} \frac{1}{\Gamma(\rho+j)} \xi^{\rho+j-1} e^{-\xi} \\
 E^j g_\rho(\xi; \delta) &= g_{\rho+j}(\xi; \delta)
 \end{aligned}$$

2.3 *The approximate formula of  $P_r\{m \operatorname{tr} L^{-1} \mathbf{V} \leq 2\xi\}$  up to the order  $n^{-1}$ .* Substituting (30) into (17), we obtain

$$\begin{aligned}
 P_r \{m \operatorname{tr} L^{-1} \mathbf{V} \leq 2\xi\} &= P_r \{m \operatorname{tr} \Lambda^{-1} \mathbf{V} \leq 2\xi\} \\
 &- \frac{1}{n} \sum_{rstu} \lambda_{ur} \lambda_{st} \left\{ \frac{m}{2} [rs|tu](E^2 + E) + \frac{m^2}{4} [rs][tu](E^2 - E) \right. \\
 &+ \frac{m}{2} [rs] \delta_{(tu)}^2 (E^3 - E^2) + \delta_{(rs, tu)}^2 E^3 + \left. \frac{1}{4} \delta_{(rs)}^2 \delta_{(tu)}^2 (E^4 - E^3) \right\} g_\rho(\xi; \rho) \\
 (31) \qquad \qquad \qquad &+ 0(n^{-2}) \\
 &= G_\rho(\xi; \rho) - \frac{1}{n} \left[ \frac{m}{4} p(p+1)(E^2 + E) + \frac{m^2}{4} p(E^2 - E) + \frac{m}{2} \delta^2 (E^3 - E^2) \right. \\
 &+ (p+1) \delta^2 E^3 + \sum_{i,j} \sum_{k,l} \lambda^{ki} \lambda^{lj} \left( \sum_{\alpha=1}^m \eta_{i\alpha} \eta_{j\alpha} \right) \left( \sum_{\alpha=1}^m \eta_{k\alpha} \eta_{l\alpha} \right) (E^4 - E^3) \left. \right] g_\rho(\xi; \delta), \\
 &+ 0(n^{-2}),
 \end{aligned}$$

since individual terms of the right hand side are easily calculated as

$$\begin{aligned}
 \sum_{rstu} \lambda_{ur} \lambda_{st} [rs|tu] &= \frac{1}{2} \sum_{rstu} \lambda_{ur} \lambda_{st} (\lambda^{ur} \lambda^{st} + \lambda^{us} \lambda^{rt}) = \frac{1}{2} p(p+1), \\
 \sum_{rstu} \lambda_{ur} \lambda_{st} [rs][tu] &= p, \\
 \sum_{rstu} \lambda_{ur} \lambda_{st} [rs] \delta_{(tu)}^2 &= \sum_{tu} \lambda_{tu} \sum_{\alpha=1}^m \eta_{\alpha} \Lambda^{-1} \Lambda_{tu} \Lambda^{-1} \eta'_{\alpha} = \sum_{\alpha=1}^m \eta_{\alpha} \Lambda^{-1} \left( \sum_{tu} \lambda_{tu} \Lambda_{tu} \right) \Lambda^{-1} \eta'_{\alpha} = \delta^2,
 \end{aligned}$$

$$\sum_{rstu} \lambda_{ur} \lambda_{st} \delta_{(rs, tu)}^2 = \sum_{rstu} \lambda_{ur} \lambda_{st} \sum_{\alpha=1}^m \eta_{\alpha} \Lambda^{-1} \Lambda_{rs} \Lambda^{-1} \Lambda_{tu} \Lambda^{-1} \eta'_{\alpha} = (p+1) \delta^2,$$

$$\sum_{rstu} \lambda_{ur} \lambda_{st} \delta_{(rs)}^2 \delta_{(tu)}^2 = \sum_{i,j} \sum_{k,l} \lambda^{ki} \lambda^{lj} \left( \sum_{\alpha=1}^m \eta_{i\alpha} \eta_{j\alpha} \right) \left( \sum_{\alpha=1}^m \eta_{k\alpha} \eta_{l\alpha} \right).$$

Then we finally obtain, as the approximation up to order  $n^{-1}$ ,

$$\begin{aligned} P_r \{ m \operatorname{tr} \mathbf{L}^{-1} \mathbf{V} \leq 2\xi \} &= G_p(\xi; \delta) - \frac{1}{n} \left[ \frac{mp}{4} (p-m+1) g_{p+1}(\xi; \delta) \right. \\ &+ \frac{m}{2} \left\{ \frac{1}{2} p(p+1) + \frac{mp}{2} - \delta^2 \right\} g_{p+2}(\xi; \delta) \\ (32) \quad &+ \left\{ \left( \frac{m}{2} + p + 1 \right) \delta^2 - \sum_{i,j} \sum_{k,l} \lambda^{ki} \lambda^{lj} \left( \sum_{\alpha=1}^m \eta_{i\alpha} \eta_{j\alpha} \right) \left( \sum_{\alpha=1}^m \eta_{k\alpha} \eta_{l\alpha} \right) \right\} g_{p+3}(\xi; \delta) \\ &\left. + \sum_{i,j} \sum_{k,l} \lambda^{ki} \lambda^{lj} \left( \sum_{\alpha=1}^m \eta_{i\alpha} \eta_{j\alpha} \right) \left( \sum_{\alpha=1}^m \eta_{k\alpha} \eta_{l\alpha} \right) g_{p+4}(\xi; \delta) \right] + 0(n^{-1}). \end{aligned}$$

The term of order  $n^{-2}$  is extremely complicated and omitted here. From (32), we can approximately evaluate the power  $P_r \{ m \operatorname{tr} \mathbf{L}^{-1} \mathbf{V} > T_0^2(\eta) \}$  under the alternative hypothesis.

It is convenient to notice that when  $p=2$ ,

$$(33) \quad \sum_{i,j} \sum_{k,h} \lambda^{ki} \lambda^{hj} \left( \sum_{\alpha=1}^m \eta_{i\alpha} \eta_{j\alpha} \right) \left( \sum_{\alpha=1}^m \eta_{k\alpha} \eta_{h\alpha} \right) = \delta^4 - 2 \left| \sum_{\alpha=1}^m \eta'_{\alpha} \eta_{\alpha} \right| / |\Lambda|$$

and when  $m=1$

$$(34) \quad \sum_{i,j} \sum_{k,l} \lambda^{ki} \lambda^{lj} \eta_i \eta_j \eta_k \eta_l = \sum_{i,j} \sum_{k,l} (\eta_i \lambda^{ki} \eta_i) (\eta_j \lambda^{lj} \eta_j) = \delta^4.$$

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