

On Elementary Symmetric Functions of the Roots
of Two Matrices in Multivariate Analysis

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1. Summary

A lemma is proved to show that the moments of the $(s-1)$ th elementary symmetric function (esf) in s non-null characteristic roots, $\lambda_i (i=1,2,\dots,s)$, of a matrix in multivariate analysis can be derived from those of the i th esf. Using this lemma the first four moments of the $(s-1)$ th esf have been obtained from those of the first esf already known (Pillai, 1954, 1960; Pillai and Samson, 1959). Further, a second lemma is given showing that the moments of the $(s-1)$ th esf in the s characteristic roots, $\theta_i = \lambda_i / (1 + \lambda_i)$, are derivable from those of the first esf in the λ 's. Upper percentage points (5% and 1%) are obtained for the distribution of the $(s-1)$ th esf in the λ 's for $s = 3$ using the moment quotients. An example is given to illustrate the use of this criterion.

2. Introduction

Many of the distribution problems in multivariate analysis are based on the distribution of the non-null characteristic roots of a matrix derived from sample observations taken from multivariate normal populations. This distribution, given by Roy (1939), is of the form

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$$f(\lambda_1, \lambda_2, \dots, \lambda_s; m, n) = C(s, m, n) \prod_{i=1}^s \left\{ \frac{\lambda_i^m}{(1+\lambda_i)^{m+n+s+1}} \right\} \prod_{i>j} (\lambda_i - \lambda_j) \quad (2.1)$$

$$0 < \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_s < \infty,$$

where

$$C(s, m, n) = \pi^{s/2} \prod_{i=1}^s \left\{ \frac{\Gamma(\frac{2m+2n+s+i+2}{2})}{\Gamma(\frac{2m+i+1}{2}) \Gamma(\frac{2n+i+1}{2}) \Gamma(i/2)} \right\} \quad (2.2)$$

and m and n are defined differently for various situations described by Pillai (1955, 1960).

The studies on the first esf in the λ 's have been carried out by Pillai (1954, 1956, 1960) and Pillai and Samson (1959). In this paper, moments of some of the other esf's are considered and in particular those of the $(s-1)$ th, $(s-2)$ th and the second esf's. Further, the esf's in the characteristic roots, $\theta_i = \lambda_i / (1 + \lambda_i)$, are studied in a similar manner.

3. Moments of the $(s-1)$ th esf in the λ 's

First consider the following lemma:

Lemma 1. If $U_{i,m,n}^{(s)}$ and $\mu_r' \{ U_{i,m,n}^{(s)} \}$ denote the i th esf in the s λ 's following the distribution (2.1), and its r th moment respectively, then

$$\mu_r' \left\{ U_{s-i,m,n}^{(s)} \right\} = \frac{C(s,m,n)}{C(s,n-r,m+r)} \mu_r' \left\{ U_{i,n-r,m+r}^{(s)} \right\}. \quad (3.1)$$

Proof. Let $\gamma_i = 1/\lambda_i$ ($i=1, 2, \dots, s$). Then the i th esf in the γ 's equals the quotient of the $(s-i)$ th esf and the s th esf in the λ 's or, reciprocally, the $(s-i)$ th esf in the λ 's equals the quotient of the i th esf and the s th esf in the γ 's. Now note that the distribution of the γ 's follows the distribution (2.1) with m and n interchanged. Hence the lemma follows.

From lemma 1, putting $i=1$ in (3.1), the r th moment of the $(s-1)$ th esf in the λ 's is obtained as

$$\mu_r' \left\{ U_{s-1, m, n}^{(s)} \right\} = \frac{c(s, m, n)}{C(s, n-r, m+r)} \mu_r' \left\{ U_{1, n-r, m+r}^{(s)} \right\}. \quad (3.2)$$

The first three central moments of $U_{1, m, n}^{(s)}$ are available in (Pillai, 1954, Pillai and Samson, 1959) and the fourth central moment in (Pillai, 1960). Using these results, the following first four moments of the $(s-1)$ th esf in the λ 's are obtained:

$$\mu_1' \left\{ U_{s-1, m, n}^{(s)} \right\} = \frac{s(2n+s-1)}{2(m+1)} \pi \sum_{i=1}^s \frac{(2m+i+1)}{(2n+i-1)}, \quad (3.3)$$

$$\mu_2' \left\{ U_{s-1, m, n}^{(s)} \right\} = \left\{ \frac{s(2n+s-3)}{4(m+2)^2} \right\} \left\{ \frac{(2m+2n+s+1)(2m+s+4)}{(m+1)(2m+5)} + s(2n+s-3) \right\} A \quad (3.4)$$

where

$$A = \pi \sum_{i=1}^s \frac{(2m+i+3)(2m+i+1)}{(2n+i-1)(2n+i-3)},$$

$$\mu_3' \left\{ U_{s-1, m, n}^{(s)} \right\} = \left\{ \frac{s(2n+s-5)}{8(m+3)^3} \right\} \left\{ \frac{(2m+2n+s+1)(2m+s+6)}{(m+2)(2m+7)} \right\} \quad (3.5)$$

$$\left\{ \frac{4(m+2n+s-2)(2m+2s+6)}{(m+1)(2m+8)} + 3s(2n+s-5) \right\} + s^2(2n+s-5)^2 \left\} B$$

where

$$B = \pi \sum_{i=1}^s \frac{(2m+i+5)(2m+i+3)(2m+i+1)}{(2n+i-1)(2n+i-3)(2n+i-5)}$$

and

$$\mu_4^* \left\{ U_{s-1, m, n}^{(s)} \right\} = \left\{ \frac{s(2n+s-7)}{16(m+4)^4} \right\} \left\{ \frac{48(2m+2n+s+1)[C+D(n-4)(m+n+s+1)]}{(m+1)(m+2)(m+3)(m+5)(2m+7)(2m+9)(2m+11)} + E \right\} F$$

..... (3.6)

where

$$C = \sum_{j=0}^6 (-1)^j f_j p^{6-j},$$

$$D = \sum_{j=0}^5 (-1)^j g_j p^{5-j},$$

$$p = (m+s+5),$$

$$E = s(2n+s-7) \left\{ \frac{2(2m+2n+s+1)(2m+s+8)}{(m+3)(2m+9)} \left[\frac{4(2n+m+s-3)(2m+2s+8)}{(m+2)(m+5)} + 3s(2n+s-7) \right] + s^2(2n+s-7)^2 \right\},$$

$$F = \prod_{i=1}^s \frac{(2m+i+7)(2m+i+5)(2m+i+3)(2m+i+1)}{(2n+i-1)(2n+i-3)(2n+i-5)(2n+i-7)}$$

and f's and g's are polynomials in s given below:

$$f_0 = s^2 + s + 4,$$

$$f_1 = 4.5s^3 + 12.5s^2 + 10s + 16,$$

$$f_2 = 8.25s^4 + 37.5s^3 + 42.5s^2 + 15.25s + 21,$$

$$f_3 = 7.875s^5 + 49.625s^4 + 93.75s^3 + 44.25s^2 - 10.25s + 7,$$

$$f_4 = 4.125s^6 + 33.1875s^5 + 89.625s^4 + 85.0625s^3 - 3.25s^2 - 34.25s - 5,$$

$$f_5 = 1.125s^7 + 11s^6 + 38.5625s^5 + 56.125s^4 + 20.9375s^3 - 24.25s^2 - 20.75s - 3,$$

$$f_6 = 0.125s^8 + 1.4375s^7 + 6.1875s^6 + 12.0625s^5 + 8.9375s^4 - 3s^3 - 7.75s^2 - 3s,$$

$$\begin{aligned}
g_0 &= 2s, \\
g_1 &= 8s^2 + 15s - 20, \\
g_2 &= 12.5s^3 + 47s^2 + 9s - 92, \\
g_3 &= 9.5s^4 + 53.75s^3 + 81s^2 - 55s - 156, \\
\text{and } g_4 &= 3.5s^5 + 26.5s^4 + 68.75s^3 + 47s^2 - 87s - 116, \\
g_5 &= 0.5s^6 + 4.75s^5 + 16.75s^4 + 26s^3 + 5s^2 - 36s - 32.
\end{aligned}$$

4. Moments of the (s-1)th esf in the θ 's

As in the previous section a lemma may be stated.

Lemma 2. If $V_{i,m,n}^{(s)}$ and $\mu_r' \left\{ V_{i,m,n}^{(s)} \right\}$ denote the i th esf in the s θ 's and its r th moment respectively, then

$$\mu_r' \left\{ V_{s-1,m,n}^{(s)} \right\} = \frac{C(s,m,n)}{C(s,n,m+r)} \sum_{j=0}^r \binom{r}{j} s^{r-j} \mu_j' \left\{ U_{1,m,m+r}^{(s)} \right\}. \quad (4.1)$$

Proof. For proving (4.1), it is enough if we note that

$$V_{s-1,m,n}^{(s)} = (s + \sum_{i=1}^s \gamma_i) / \prod_{i=1}^s (1 + \gamma_i) \quad \text{where } \gamma_i \text{'s} \quad (4.2)$$

as stated above, follow the distribution (2.1) with m and n interchanged. The rest easily follows.

The first four moments of the (s-1)th esf in the θ 's thus may be obtained by using lemma 2. However they are not given here since it has been shown (Pillai, 1964) that the moments of $V_{i,m,n}^{(s)}$ can be derived in a simple manner from the respective moments of $U_{i,m,n}^{(s)}$ (and vice versa).

5. Moments of other esf's in the θ 's and λ 's

It has been shown (Pillai, 1954, 1956; Pillai and Mijares, 1959) that the distribution of θ 's obtainable from (2.1) by the transformation $\lambda_i = \theta_i / (1 - \theta_i) (i=1, 2, \dots, s)$ can be expressed in a determinantal form and when integrated within the limits, $0 < \theta_1 \leq \dots \leq \theta_s < 1$, is given by

$$C(s, m, n) V(s-1, s-2, \dots, 1, 0) = C(s, m, n) \begin{vmatrix} \int_0^1 \theta_s^{m+s-1} (1-\theta_s)^n d\theta_s & \dots & \int_0^1 \theta_s^m (1-\theta_s)^n d\theta_s \\ \dots & \dots & \dots \\ \int_0^{\theta_2} \theta_1^{m+s-1} (1-\theta_1)^n d\theta_1 & \dots & \int_0^{\theta_2} \theta_1^m (1-\theta_1)^n d\theta_1 \end{vmatrix} \quad (5.1)$$

Further,

$$\mu_1^i \left\{ V_{i, m, n}^{(s)} \right\} = C(s, m, n) V(s, s-1, \dots, s-i+1, s-i-1, \dots, 1, 0). \quad (5.2)$$

The right hand side of (5.2) can be shown to be equal to

$$\binom{s}{i} \pi \prod_{j=1}^i \frac{(2m+s-j+2)}{(2m+2n+2s-j+3)} \quad (5.3)$$

based on some particular cases evaluated by Pillai and Mijares (1959). From this result, using the method given by Pillai (1964) i.e. by attaching negative signs to all terms except that in n in each linear factor involving n in (5.3) and further changing n to $m+n+s+1$, the first moment of the i th esf in the λ 's is given by

$$\mu_1^i \left\{ U_{i, m, n}^{(s)} \right\} = \binom{s}{i} \pi \prod_{j=1}^i \frac{(2m+s-j+2)}{(2n+j-1)}. \quad (5.4)$$

Now consider $\mu_2' \left\{ V_{2,m,n}^{(s)} \right\}$. It can be shown that (Pillai (1964))

$$\begin{aligned} \mu_2' \left\{ V_{2,m,n}^{(s)} \right\} = C(s,m,n) & \left[V(s+1,s,s-3,\dots,1,0) + V(s+1,s-1,s-2,s-4,\dots,1,0) \right. \\ & \left. + V(s,s-1,s-2,s-3,s-5,s-6,\dots,1,0) \right] . \end{aligned} \quad \dots(5.5)$$

Now substituting the values of the determinants in (5.5), already evaluated by Pillai and Mijares (1959),

$$\mu_2' \left\{ V_{2,m,n}^{(s)} \right\} = \frac{s(s-1)(2m+s)(2m+s+1)}{6} G_1 \quad (5.6)$$

$$3! \prod_{j=1}^{2m+2n+2s-j+5} (2m+2n+2s-j+5)$$

$$\begin{aligned} \text{where } G_1 = & 6n^2 \left\{ 4s(s-1)m^2 + 2(s-1)(2s^2+s+8)m + s^4 + 7s^2 - 8s + 12 \right\} \\ & + 3n \left\{ 16s(s-1)m^3 + 4(s-1)(8s^2+5s+8)m^2 + 2(s-1)(10s^3+12s^2+27s+24)m \right. \\ & \left. + 4s^5 + 3s^4 + 12s^3 + 5s^2 - 24s + 36 \right\} + s(s+1)(2m+s+1)(2m+s+2)(m+s)(2m+2s+1) \\ & + (s-2)(2m+2s+3)(2m+s-1) \left\{ 4sm^2 + 2s(3s+2)m + 2s^3 + 3s^2 + s + 6 \right\} . \end{aligned}$$

The second moment of the second esf in the λ 's can be obtained from (5.6) using the method given by Pillai (1964). We get

$$\mu_2' \left\{ U_{2,m,n}^{(s)} \right\} = \frac{s(s-1)}{3!} \frac{(2m+s)(2m+s+1)}{6} \cdot G \quad (5.7)$$

$$\prod_{j=1}^{2n+j-3} (2n+j-3)$$

where G is obtained from G_1 by attaching negative sign to the first degree term in n and then changing n to $m+n+s+1$ in all the terms.

Further, applying (3.1) with $i=2$ and $r=2$, we get from (5.7)

$$\mu_2^i \left\{ U_{s-2,m,n}^{(s)} \right\} = \frac{C(s,m,n)}{C(s,n-2,m+2)} \mu_2^i \left\{ U_{2,n-2,m+2}^{(s)} \right\}. \quad (5.8)$$

6. Upper percentage points of $U_{2,m,n}^{(3)}$

Using (3.3) to (3.6) with $s=3$, the moment quotients, β_1 and β_2 , were computed on IBM 7090 accurate to five decimals for values of $m = -\frac{1}{2} \left(\frac{1}{2}\right) 5, 7, 10(5) 50, 60, 80, 100, 130, 160, 200, 300, 500$ and 1000 and n starting from 20 but otherwise as for m . The upper percentage points were computed manually using tables of "Percentage points of Pearson curves for given β_1 , β_2 expressed in standardized measure" (Pearson and Hartley, 1958) and extrapolating in some cases from "Tables of deviates of Type I curves measured from the mean in terms of the standard deviation" (Pearson, 1931). Tables 1 and 2 give the percentage points thus computed.

7. An Example

The use of the percentage points for $U_{2,m,n}^{(3)}$ may be illustrated by an example. This criterion is being suggested for testing different kinds of hypotheses in multivariate analysis, for example, i) that of equality of the dispersion matrices of two p -variate normal populations, ii) that of equality of the p -dimensional mean vectors for ℓ p -variate normal populations and, iii) that of independence between a p -set and a q -set of variates in a $(p+q)$ -variate normal population. It may be pointed out that for the last hypothesis, transformation has to be carried out from θ 's to the λ 's since computationally it is easier to obtain θ 's first. Now, for illustration, hypothesis ii) is considered here using the following data taken from Rao (1952, p. 263) concerning measurements of 140 schoolboys of almost the same age

belonging to six different schools in an Indian city. Three characters were measured (1) head length, (2) height and (3) weight. The problem is to test for the significance of differences between schools in the mean characters. Let S^* , S be the matrices consisting of the sums of squares and products 'between' and 'within' schools for the three characters. The S^* and S matrices as well as the values of $m = 0.5$ and $n = 65$ are given in Pillai and Samson (1959). Using those matrices, S^{-1} , the inverse of S , and $S^* S^{-1}$ are given below:

$$S^{-1} = \begin{pmatrix} .0^4 823,947,419 & & \\ -.0^4 572,276,500 & .0^2 148,849,391 & \\ .0^6 756,373,421 & -.0^3 284,873,914 & .0^3 103,413,946 \end{pmatrix} \quad (7.1)$$

and

$$S^* S^{-1} = \begin{pmatrix} .050,096,98 & .127,295,4 & -.006,541,510 \\ .009,293,867 & .098,659,55 & -.001,449,733 \\ .021,212,45 & .107,934,8 & .052,878,55 \end{pmatrix} \quad (7.2)$$

From (7.2), $U_{2,0.5,65}^{(3)} = .01192$. By interpolation from Tables 1 and 2 respectively it may be seen that this value is significant at the upper 5% level but not significant at the upper 1% level. This agrees with the findings (see Pillai and Samson, 1959) based on the test $U_{1,0.5,65}^{(3)}$ and also with those of Rao, who examined the data using the Λ criterion of Pearson and Wilks (1933). Foster (1957), however, finds that the largest root is not significant at the upper 5% level but only at the upper 15% level.

It may be pointed out that the monotonicity of the power function of this test has been already established (Das Gupta, Anderson and Mudholkar, 1964).

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Table 1. Upper 5% Points of $U_{2,m,n}^{(3)}$

$\frac{n}{m}$	20	25	30	35	40	45	50	60	80	100	130	160	200	300	500	1000
.5	.036	.023	.016	.012	.0089	.0070	.0057	.0040	.0022	.00142	.00084	.00055	.00035	.000157	.000057	.0000142
0	.064	.041	.028	.021	.016	.012	.009	.0070	.0039	.00250	.00148	.00097	.00062	.000277	.000100	.0000250
1	.098	.062	.043	.032	.024	.019	.015	.011	.0060	.00383	.00226	.00149	.00095	.00042	.000153	.0000382
1.5	.139	.088	.061	.045	.034	.027	.022	.015	.0085	.00540	.00318	.00211	.00135	.000795	.000286	.0000714
2	.186	.118	.081	.060	.045	.036	.028	.020	.0112	.00719	.00424	.00281	.00179	.00101	.000365	.0000911
2.5	.237	.151	.104	.076	.058	.046	.037	.026	.0144	.00918	.00542	.00358	.00229	.00126	.000452	.000113
3	.292	.188	.130	.095	.072	.057	.046	.032	.0178	.0114	.00673	.00443	.00283	.00152	.0004948	.000136
3.5	.352	.213	.149	.105	.078	.063	.051	.039	.0215	.0138	.00815	.00537	.00343	.00181	.000548	.000162
4	.422	.259	.181	.126	.093	.072	.057	.044	.0258	.0164	.00972	.00640	.00409	.00213	.000763	.000190
4.5	.509	.321	.222	.162	.113	.089	.071	.054	.0302	.0193	.0114	.00750	.00479	.00246	.000882	.000220
5	.618	.372	.255	.187	.133	.104	.086	.063	.0350	.0223	.0132	.00868	.00534	.00281	.00101	.000252
7	.800	.468	.311	.225	.164	.129	.104	.072	.0400	.0255	.0151	.00992	.00634	.00344	.00129	.000358
10	1.08	.681	.468	.341	.243	.185	.143	.104	.0634	.0404	.0238	.0157	.0100	.00444	.00179	.000473
15	1.86	1.17	.800	.582	.443	.348	.281	.206	.108	.0687	.0404	.0256	.0170	.00752	.00270	.000627
20	3.6	2.23	1.54	1.12	.850	.667	.537	.401	.206	.131	.0770	.0507	.0323	.0231	.00827	.00206
25	8.7	5.44	3.71	2.69	2.04	1.60	1.28	.985	.534	.312	.185	.120	.0765	.0338	.0121	.00301
30	12.1	7.54	5.14	3.73	2.82	2.21	1.76	1.32	.78	.430	.252	.165	.105	.0454	.0166	.00344
35	16.0	9.98	6.79	4.93	3.73	2.92	2.35	1.61	.893	.566	.332	.217	.139	.0611	.0278	.00544
40	20.5	12.76	8.69	6.29	4.76	3.73	3.00	2.06	1.14	.720	.422	.277	.176	.0777	.0344	.00855
45	25.6	15.87	10.80	7.82	5.92	4.63	3.72	2.55	1.41	.892	.523	.343	.218	.0962	.0417	.0104
50	31	19.33	13.45	9.51	7.20	5.63	4.52	3.10	1.71	1.08	.634	.416	.265	.117	.0582	.0145
60	44	27.2	18.53	13.4	10.1	7.92	6.36	4.35	2.40	1.28	.889	.583	.370	.163	.082	.0247
80	76	47.1	32.0	23.1	17.5	13.7	11.0	7.49	4.13	1.82	1.52	.998	.634	.279	.0991	.0372
100	117	72.4	49.2	35.5	26.8	20.9	16.8	11.47	6.31	2.61	2.33	1.52	.966	.424	.151	.0616
130	195	120	81.7	59.0	44.2	34.7	27.8	19.0	10.5	3.99	3.84	2.51	1.593	.699	.249	.0916
150	292	181	114.2	88.4	66.6	52.0	41.7	28.4	15.6	6.59	5.74	3.75	2.38	1.04	.370	.141
200	452	280	166.6	137	103	80.4	64.4	43.9	24.1	9.85	8.84	5.77	3.65	1.60	.569	.308
300	1000	622	421	304	229	178	143	97.3	53.4	15.1	19.5	12.74	8.06	3.52	1.249	.836
500	2766	1712	1159	835	629	490	392	267	116	32.0	53.4	34.6	22.00	9.59	3.39	.836
1000	10991	6797	4599	3310	2493	1941	1553	1058	579	362	211	137	86.7	37.1	13.32	3.271

Table 2. Upper 1% Points of $V_{2,m,n}^{(3)}$

$m \backslash n$	20	25	30	35	40	45	50	60	80	100	130	160	200	300	500	1000
.5	.077	.051	.034	.025	.019	.015	.012	.0082	.0045	.0029	.0017	.0011	.00071	.00031	.00011	.00028
0	.12	.076	.052	.037	.028	.022	.018	.012	.0069	.0043	.0026	.0017	.0011	.00048	.00017	.00043
1	.17	.11	.072	.052	.040	.031	.025	.017	.0097	.0061	.0036	.0024	.0015	.00067	.00024	.00061
1.5	.23	.14	.096	.070	.053	.041	.033	.023	.013	.0083	.0047	.0032	.0020	.00090	.00032	.00080
2	.29	.18	.12	.089	.068	.053	.043	.030	.017	.010	.0062	.0041	.0026	.0014	.00041	.00102
2.5	.36	.22	.15	.11	.084	.066	.053	.037	.021	.013	.0077	.0051	.0032	.0018	.00051	.00155
3	.44	.27	.19	.14	.10	.081	.065	.045	.025	.016	.0093	.0061	.0039	.0021	.00074	.00185
3.5	.52	.33	.22	.16	.12	.096	.077	.053	.030	.019	.0112	.0073	.0047	.0024	.00087	.00217
4	.62	.38	.26	.19	.14	.113	.091	.063	.035	.022	.0131	.0086	.0055	.00281	.00101	.00252
4.5	.72	.44	.30	.22	.17	.132	.106	.073	.041	.026	.0151	.0100	.0064	.00324	.00115	.00289
5	.83	.51	.35	.25	.19	.151	.121	.083	.047	.030	.0174	.0114	.0073	.00367	.00131	.00328
5.5	.94	.58	.40	.29	.22	.172	.138	.095	.053	.034	.0197	.0130	.0083	.00367	.00203	.00505
6	1.06	.66	.46	.34	.26	.206	.164	.114	.066	.042	.022	.015	.010	.00367	.00233	.00829
7	1.21	.75	.52	.39	.30	.236	.188	.134	.075	.048	.025	.017	.012	.00367	.00263	.00970
10	1.46	1.01	0.73	0.53	0.40	0.30	0.23	0.16	0.09	0.055	0.028	0.018	0.012	0.00367	0.00270	0.0153
15	1.75	1.21	0.87	0.63	0.48	0.36	0.28	0.19	0.11	0.069	0.035	0.021	0.014	0.00367	0.00270	0.0241
20	2.04	1.43	1.01	0.73	0.54	0.40	0.31	0.21	0.12	0.075	0.038	0.024	0.016	0.00367	0.00270	0.0348
25	2.31	1.63	1.14	0.82	0.59	0.43	0.33	0.23	0.13	0.082	0.041	0.026	0.018	0.00367	0.00270	0.0473
30	2.56	1.81	1.26	0.90	0.64	0.46	0.35	0.25	0.14	0.091	0.046	0.029	0.020	0.00367	0.00270	0.0616
35	2.81	2.00	1.40	1.00	0.71	0.51	0.39	0.28	0.16	0.102	0.051	0.032	0.022	0.00367	0.00270	0.0777
40	3.06	2.19	1.54	1.10	0.77	0.55	0.42	0.30	0.18	0.114	0.056	0.035	0.024	0.00367	0.00270	0.0956
45	3.31	2.38	1.68	1.20	0.83	0.59	0.45	0.32	0.19	0.126	0.061	0.038	0.026	0.00367	0.00270	0.115
50	3.56	2.57	1.81	1.30	0.89	0.63	0.48	0.34	0.20	0.138	0.066	0.041	0.028	0.00367	0.00270	0.130
60	4.06	2.96	2.07	1.47	0.99	0.69	0.52	0.37	0.22	0.150	0.071	0.044	0.030	0.00367	0.00270	0.160
80	4.81	3.46	2.40	1.70	1.14	0.80	0.59	0.41	0.25	0.162	0.076	0.047	0.032	0.00367	0.00270	0.200
100	5.56	3.96	2.73	1.91	1.26	0.87	0.63	0.44	0.27	0.174	0.081	0.050	0.034	0.00367	0.00270	0.247
130	6.31	4.46	3.06	2.12	1.38	0.94	0.67	0.47	0.29	0.186	0.086	0.053	0.036	0.00367	0.00270	0.298
160	7.06	4.96	3.39	2.33	1.50	1.01	0.70	0.50	0.31	0.198	0.091	0.056	0.038	0.00367	0.00270	0.349
200	8.06	5.46	3.72	2.54	1.62	1.08	0.75	0.53	0.33	0.210	0.096	0.059	0.040	0.00367	0.00270	0.400
300	9.56	6.46	4.35	2.96	1.84	1.18	0.81	0.58	0.36	0.222	0.101	0.062	0.042	0.00367	0.00270	0.460
500	12.56	8.46	5.64	3.77	2.26	1.38	0.94	0.63	0.40	0.234	0.106	0.065	0.044	0.00367	0.00270	0.520
1000	13.351	8.076	5.377	3.625	2.051	1.205	0.753	0.511	0.317	0.246	0.111	0.068	0.046	0.00367	0.00270	0.580