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MULTIVARIATE F MATRIX

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# Title

**On the Limiting Empirical Distribution Function of the Eigenvalues of a Multivariate F Matrix**

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- Canonical correlation analysis; empirical distribution function; large dimensional random matrices; limiting distribution; multivariate analysis of variance; (CONT.)

## Abstract

In this paper, the authors derived an explicit expression for the limit of the empirical distribution function of a central multivariate F matrix when the number of variables and degrees of freedom tend to infinity in certain fashion. This distribution is useful in deriving the limiting distributions of certain test statistics which arise in multivariate analysis of variance, canonical correlation analysis and tests for the equality of two covariance matrices.

**ITEM #18, SUBJECT TERMS, CONTINUED:** multivariate F matrix.

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ON LIMITING EMPIRICAL DISTRIBUTION FUNCTION OF
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Z. D. Bai, Y. Q. Yin and P. R. Krishnaiah

ABSTRACT

In this paper, the authors derived an explicit expression for the limit of the empirical distribution function of a central multivariate F matrix when the number of variables and degrees of freedom tend to infinity in certain fashion. This distribution is useful in deriving the limiting distributions of certain test statistics which arise in multivariate analysis of variance, canonical correlation analysis and tests for the equality of two covariance matrices.

Keywords and Phrases: Canonical correlation analysis, empirical distribution function, large dimensional random matrices, limiting distribution, multivariate analysis of variance, multivariate F matrix.
1. INTRODUCTION

Various test procedures in multivariate analysis are based upon certain functions of the eigenvalues of random matrices. A considerable amount of work was done in the literature on the asymptotic distribution theory of these statistics when the sample size is very large. But, many situations arise in multivariate data analysis when the number of variables and the sample size are both very large. So, there is a great need to investigate the distributions of various functions of the eigenvalues of large dimensional random matrices. Distributions of the eigenvalues of large dimensional random matrices arise (e.g., see Mehta (1967)) in nuclear physics also.

Some work was done in the literature on the limiting empirical distribution function (e.d.f.) of large dimensional random matrices. Here, we note that the e.d.f. of a random matrix \( Z: p \times p \) is defined as \( N(x)/p \) where \( N(x) \) denotes the number of the eigenvalues of \( Z \) which are less than or equal to \( x \). The e.d.f. (also known as spectral distribution) of \( Z \) is useful in deriving the distributions of certain functions of the eigenvalues of \( Z \).

Now, let \( S_1: p \times p \) be distributed as central Wishart matrix with \( m \) degrees of freedom and \( E(S_1/m) = I_p \). Also, let \( p \) and \( m \) both tend to infinity such that \( \lim(p/m) = y > 0 \). Then, it is known (see Grenander and Silverstein (1977), Jonsson (1982) and Wachter (1978)) that the e.d.f. \( F_m(x) \) of the eigenvalues of \( S_1/m \) tends to \( F_y(x) \) where \( F_y(x) \) is a distribution function with density function given by

\[
f_y(x) = \begin{cases} \frac{\sqrt{(x-a)(b-x)}}{2\pi y} & a < x < b \\ 0 & \text{otherwise} \end{cases}
\]

where \( a = (1 - \sqrt{y})^2 \), and \( b = (1 + \sqrt{y})^2 \) and \( 0 < y \leq 1 \); for \( 1 < y < \infty \), \( F_y(x) \) has mass \( 1 - (1/y) \) at zero and \( f_y(x) \) on \((a,b)\). Yin and Krishnaiah (1983b) showed that the spectral distribution of the sample covariance matrix has a limit when the underlying distribution is isotropic and \( y < 1 \).
Yin and Krishnaiah (1983a) showed that the spectral distribution of $S_1 T / m$ tends to a limit in probability for each $x$ under the following conditions:

(a) $T$ is a symmetric, positive definite matrix and $C(x)$ is the e.d.f. of the eigenvalues of $T$,

(b) $S_1$ and $T$ are independent of each other,

(c) $\lim (p/m) = y$ exists and finite $p, m \to \infty$

(d) $\int x^k dG(x) \to H_k$ exists in $L^2$ for $k = 1, 2, \ldots$ and $\sum k^2 H_k = \infty$.

Yin and Krishnaiah (1984b) extended the above result to the case when $S_1$ is the sample sums of squares and cross products matrix based upon observations from an isotropic population.

Now, let $S_1: p \times p$ and $S_2: p \times p$ be distributed independently as central Wishart matrices with $m$ and $n$ degrees of freedom and $E(S_1/m) = E(S_2/n) = I_p$. Then, the distribution of $n S_1 S_2^{-1} / m$ is known to be the central multivariate $F$ matrix. Applying the result of Yin and Krishnaiah (1983), Yin, Bai, and Krishnaiah (1983) showed that the limit of the spectral distribution of the central multivariate $F$ matrix exists when $p/m \to y'$ exists and $(p/n) \to y < \frac{1}{2}$ as $p \to \infty$. Silverstein (1984b) showed the validity of the above result even for the case $\frac{1}{2} < y < 1$ by making a minor modification in the proof of Yin, Bai, and Krishnaiah (1983). Yin and Krishnaiah (1983a) gave an expression for the moments of the limit of the e.d.f. of the eigenvalues of $S_1 T / m$. Starting from this expression, Silverstein (1984b) derived an explicit expression for the limit of the e.d.f. of the eigenvalues of the multivariate $F$ matrix. In this paper, the authors give an alternative derivation of the above limit. The authors also gave explicit expressions for the moments of the above limiting distribution and these expressions are not known in the literature. Extensions of the results of this paper for nonnormal populations and nonnull cases are under investigation.
2. PRELIMINARIES

In this section, we give some results which are needed in the sequel as well as a brief review of known results on limiting spectral distribution of a multivariate F matrix.

Lemma 2.1 Let \( z \in (0,1) \), \( a' = (1 - \sqrt{z})^2 \) and \( b' = (1 + \sqrt{z})^2 \). If \( 0 < |t| < a' \), then

\[
\frac{1}{2\pi z} \int_{a'}^{b'} \frac{1}{(x - t)} \left( (x - a')(b' - x) \right)^{1/2} dx = \frac{1}{2z} [1 + z - t - ((1 - z - t)^2 - 4tz)^{1/2}] 
\]

Proof Making the transformation \( u = \frac{2x - (b' + a')}{(b' - a')} \) in the left side of (2.1), we obtain

\[
R(t) = \frac{1}{2\pi z} \int_{a'}^{b'} \frac{1}{(x - t)} \left( (x - a')(b' - x) \right)^{1/2} dx = \frac{(b' - a')}{4\pi z} \int_{-1}^{1} \frac{(1 - u^2)^{1/2}}{u + \Delta} \, du 
\]

where \( \Delta = (b' + a' - 2t)/(b' - a') \). It is known (see Jonsson (1982)) that \( R(0) = 1 \). So, for any \( r \in (0,1) \), we have

\[
\frac{1}{\sqrt{r}} \int_{-1}^{1} \frac{(1 - u^2)^{1/2}}{u + ((1 + r)/2\sqrt{r})} = 1. \tag{2.3}
\]

Now, let \( \Delta = (1 + r)/2\sqrt{r} \). Since \( \Delta > 1 \), the condition \( r \in (0,1) \) is satisfied.

So, using (2.2) and (2.3), we obtain

\[
R(t) = (b' - a')(\Delta - \sqrt{\Delta^2 - 1})/4z = \frac{1 + z - t - ((1 - z - t)^2 - 4tz)^{1/2}}{2z}. \tag{2.4}
\]

Lemma 2.2 For any nonnegative integers \( m \) and \( w \), we have

\[
\sum_{t=0}^{[w/2]} (-1)^t \binom{m}{t} \binom{2m - 2t}{w} = 2^m \binom{m}{w}. \tag{2.5}
\]
Here \( \binom{m}{w} \) is defined to be zero when \( m < w \).

Proof If \( m = 0 \), the proof is trivial. We now prove the result for \( m > 0 \) by induction. Suppose the result is true for a fixed value of \( m \). Then, we have

\[
\frac{[w/2]}{\sum_{k=0}^{\lfloor w/2 \rfloor} (-1)^k \binom{m+1}{k} (2m + 2 - 2k)} = \sum_{k=0}^{m+1} (-1)^k \left[ \binom{m}{k} + \binom{m}{k-1} \right] (2m + 2 - 2k)
\]

\[
= \sum_{k=0}^{m} (-1)^k \binom{m}{k} \left[ \binom{2m - 2k}{2m - 2k} + 2 \binom{2m - 2k}{2m + 1 - w} \right]
\]

\[
= 2^m \binom{m}{w} + 2 \cdot 2^{m-1} \binom{m}{w-1} = 2^w \binom{m+1}{w}.
\]

So the result follows.

Lemma 3

\[
\int \frac{(\alpha y + \beta)dy}{(\gamma + \delta y - \rho y^2)^{3/2}} = \frac{\delta y - 2\delta \gamma + (4\rho \beta + 2\rho \delta) \gamma}{(\delta^2 + 4\gamma \rho) \sqrt{\gamma + \delta y - \rho y^2}} + \text{const.}
\]

The above lemma can be verified directly by differentiation.
3. MOMENTS OF THE LIMITING SPECTRAL DISTRIBUTION
OF A MULTIVARIATE F MATRIX

Let \( F_{y,z}(x) \) denote the limit of the e.d.f. of the multivariate F matrix as defined in the preceding section. Also, let \( \{E_k\}_{k=1}^\infty \) denote the moments of \( F_{y,z}(x) \). Then, from Yin and Krishnaiah (1983), we know that

\[
E_k = \sum_{w=0}^{k-1} \frac{1}{w+1} B(k,w) \tag{3.1}
\]

where

\[
B(k,w) = \sum_{n_1,\ldots,n_{k-w}} \frac{(w+1)!}{n_1! \cdots n_{k-w}!} H_1 \cdots H_{k-w} \tag{3.2}
\]

and the summation in (3.2) is over all possible values of \( n_1,\ldots,n_{k-w} \) subject to the restrictions \( n_1 + \cdots + n_{k-w} = w + 1 \) and \( n_1 + n_2 + \cdots + (k - w)n_{k-w} = k \).

Also, \( H_i = E(x^{-i}) \) for \( i = 1, 2, \ldots \) where the density of \( x \) is given by

\[
g_z(x) = \begin{cases} \frac{(x - a')(b' - x)}{2\pi xz} & a' < x < b' \\ 0 & \text{otherwise} \end{cases} \tag{3.3}
\]

where \( a', b' \), and \( z \) are as defined in the preceding section. For any \( |t| < a' \),

\[
\phi(t) = t^1H_1 + t^2H_2 + t^3H_3 + \ldots
\]

\[
= E(t/(x - t))
\]

\[
= \frac{1}{2\pi z} \int_{a'}^{b'} [t((x - a')(b' - x))^{1/2}/(x - t)] dx
\]

\[
= (1 - z - t)[1 - ((4tz/(1 - z - t)^2))^{1/2}]/2z \tag{3.4}
\]

by using Lemma 2.1. Since \( B(k,w) \) is the coefficient of \( t^k \) in Taylor expansion of \( \phi(t)^{w+1} \), we obtain

\[
B(k,w) = \sum_{\ell=0}^{w+1} \binom{w+1}{\ell} \frac{(-1)^\ell}{\ell!} \sum_{j=0}^{\ell/2} \binom{\ell}{j} (-1)^j 2^j - 1 z^{j-w-1}
\]

\[
\cdot \binom{w + 1 - 2j}{k - j} (1 - z)^{w+1-j-k} (-1)^{k-j} \tag{3.5}
\]
If \( w + 1 - 2j \geq 0 \), \( \left( \begin{array}{c} w + 1 - 2j \\ k - j \end{array} \right) \neq 0 \) implies \( 0 \leq j \leq w + 1 - k \leq 0 \), i.e. \( j = 0 \).

Noticing \( \sum_{\xi=0}^{w+1} \xi^k \xi! = 0 \), we see that the expression of \( B(k,w) \) only contains the terms with \( j > \frac{w + 1}{2} \). If \( \xi \) is even, then \( \left( \frac{\xi}{2} \right)! \neq 0 \) implies \( j \leq \frac{\xi}{2} \leq \frac{w + 1}{2} \), which is contrary to \( j > \frac{w + 1}{2} \). Thus, in the expression of \( B(k,w) \), there are only the terms with \( j > \frac{w + 1}{2} \) and \( \xi \) being odd. Applying lemma 2.2, we obtain

\[
B(k,w) = \sum_{\xi=0}^{w+1} \left( \begin{array}{c} w + 1 \\ 2\xi + 1 \end{array} \right) (-1)^{k+1} \sum_{j=\left[\frac{w+3}{2}\right]}^{(w+1)/2} \left( \begin{array}{c} \xi + \frac{1}{2} \\ j \end{array} \right) z^{2\xi-1-w-j-w-1}

\times \left( \begin{array}{c} w + 1 - 2j \\ k - j \end{array} \right) (1 - z)^{w+1-j-k}
\]

\[
= \sum_{\xi=0}^{w+1} \left( \begin{array}{c} w + 1 \\ 2\xi + 1 \end{array} \right) \left( \begin{array}{c} (2\xi + 1)! (2\xi - 2 - 2\xi)! \\ j! \xi! (j - \xi - 1)! \end{array} \right)
\times (-1)^{\xi+1} z^{2w-2j-w-1}(1 - z)^{w+1-j-k}
\left( \begin{array}{c} k + j - w - 2 \\ k - j \end{array} \right)
\]

\[
= \sum_{j=w+1}^{k-1} \frac{(k + j - w - 2)! (w + 1)!}{j! (j + w + 1)! (k - j)!} z^j (1 - z)^{-k-j}
\]

From (3.5) and (3.6) it follows that

\[
e_k^w = \sum_{w=0}^{k-1} \left( \begin{array}{c} k \\ w \end{array} \right) \sum_{j=0}^{k-w-1} \frac{(k + j - 1)!}{j! (j + w + 1)! (k - j - w - 1)!} z^j (1 - z)^{-k-j}
\]

\[
= \sum_{j=0}^{k-1} z^j (1 - z)^{-k-j} \sum_{w=0}^{k-1-j} \left( \begin{array}{c} k \\ w \end{array} \right) \frac{(k + j - 1)!}{j! (j + w + 1)! (k - j - w - 1)!} .
\]

The \( k \)-th moment of \( \ell \) can be easily computed from (3.7)
4. DERIVATION OF THE LIMITING SPECTRAL DISTRIBUTION OF MULTIVARIATE F MATRIX

Substituting the well-known formulas

\[
\begin{pmatrix} k \\ w \end{pmatrix} = \sum_{t=0}^{\infty} \begin{pmatrix} k - j - w - 1 \\ t \\ j + w + 1 \end{pmatrix} \left( \begin{array}{c} w - t \\ \end{array} \right) \quad (4.1)
\]

into (3.7), we obtain the following by changing the order of summation:

\[
E_k = \sum_{j=0}^{k-1} z^j (1 - z)^{-k-j} \sum_{t=0}^{\infty} \frac{(k + j - 1)! \cdot (1 + y)^{k-j-1-2t}}{(k - j - 1 - 2t)! \cdot t!(j + t + 1)!} y^t (1 + y)^{k-j-1-2t} (4.2)
\]

Using the formulas

\[
\begin{pmatrix} k + j - 1 \\ 2t \end{pmatrix} = \sum_{s=0}^{2t} \begin{pmatrix} k - 1 \\ s \end{pmatrix} \begin{pmatrix} j \\ 2t - s \end{pmatrix} \quad (4.3)
\]

in (4.2), we obtain

\[
E_k = \sum_{t=0}^{k-1} \left( \begin{array}{c} 2t \\ t + 1 \end{array} \right) \left( \begin{array}{c} k - 1 \\ s \\ j \\ \end{array} \right) \sum_{t=0}^{s-1} \frac{z^j (1 - z)^{-k-j}}{s!} y^t (1 + y)^{k-j-1-2t} (1 - z)^{t-j}(1 + y)^{k-j-1-2t} (4.4)
\]
\[
= \sum_{t=0}^{k-1} \binom{2t}{t} \frac{1}{t+1} \sum_{s=0}^{k-1} \left( \binom{k-1}{t} \right) \left( \binom{s}{s} \right) z^{-s} (1-z)^{-k-t} x (1+y)^{k-t-s-1} (y+z)^{s} \]  
(4.4)

\[
= \sum_{s=0}^{(k-1)/2} \sum_{t=s}^{k-1} \binom{k-1-s}{s+t} \binom{2t}{t} \frac{1}{t+1} z^{-s} (1-z)^{-k-t} x (1+y)^{k-s-1-t} (y+z)^{s} \]  
(4.5)

Define a random vector \((U, V)\) where the marginal density of \(U\) is \(\frac{2}{\pi} (1-x^2)^{-k/2} I_{(0,1)}(x)\) and the conditional density of \(V\) given \(U = x\) is

\[
\frac{|y|}{1-x^2} I_{[-\sqrt{1-x^2}, \sqrt{1-x^2}]} (y). \]  
(4.6)

It is easy to see that

\[
E\{U^{2s} V^{2t+2s}\} = \frac{(2s)! (2t+2s)!}{s!(2s+t)! (s+t+1)!} \frac{4^{-2s-t}}{4^{2s+t}} \]  
(4.7)

and that

\[
E\{U^{2s+1} V^{2t+2s+1}\} = 0. \]

Hence, from (4.5), (4.7) it follows that

\[
E_k = E \sum_{s=0}^{(k-1)/2} \sum_{t=s}^{k-1} \left( \binom{k-1-s}{s} \right) \left( \binom{k-1}{t} \right) z^{-s} (1-z)^{-k-t-s/2} x (1+y)^{k-s-1-t} (y+z)^{s} \]  

\[
= \frac{1}{1-z} E \left( 1 + y + \frac{4V^2 z}{1-z} + \frac{4UV \sqrt{y+z}}{\sqrt{1-z}} \right)^{k-1} \]  
(4.8)

Now, we compute the distribution of...
Let $w_1 = UV$ and $w_2 = V^2$. Then we can easily show that the joint distribution of $(w_1, w_2)$ is
\[
\frac{1}{\pi w_2} (w_2 - w_1^2)^{-3/2} \prod_{[w_1 + w_2^2 < w_2]} I
\]
where $I_{[a < b]}$ takes value 1 if $a < b$ is true and zero otherwise. Applying Lemma 2.3, we can compute the density of $\frac{4\sqrt{y + z}}{1 - z} w_1 + \frac{4z}{1 - z} w_2$ as follows:

Let $q_1 < q_2$ be the two roots of the equation
\[
\frac{(1 - z)^2}{16z^2} q^2 + \frac{1 - z}{16(y + z)} (x - q)^2 = \frac{1 - z}{4z} q
\]
in the variable $q$. Let $\alpha = 1$, $\beta = 0$, $r = \{(1 - z)/16(y + z)\}^2$, $\delta = \frac{1 - z}{4z} + \frac{1 - z}{8(y + z)} x$ and $\rho = \frac{1 - z}{16(y + z)}$. Then
\[
\begin{align*}
\Delta_1 &= 4\alpha r - 2\beta \delta = - \frac{1 - z}{4(y + z)} x^2, \\
\Delta_2 &= 4\beta \rho + 2\alpha \delta = \frac{1 - z}{2z} + \frac{1 - z}{4(y + z)} x \\
\Delta_3 &= \delta^2 + 4\gamma \rho = \frac{(1 - z)^2}{16z^2} (y + z + xz) \\
\end{align*}
\]
By Lemma 2.3, we get the density of $\frac{4\sqrt{y + z}}{1 - z} w_1 + \frac{4z}{1 - z} w_2$ as given below:

\[
f_1(x) = \frac{1}{\pi} \frac{\sqrt{1 - z}}{4\sqrt{y + z}} \frac{(1 - z)^2}{16z^2} \int_{q_1}^{q_2} q \left( \frac{1 - z}{4z} q - \frac{1 - z}{16(y + z)} (x - q)^2 \right)^{-3/2} dq
\]
\[
= \frac{(1 - z)^{5/2}}{64\pi z^2\sqrt{y + z}} \frac{\Delta_1 + \Delta_2 q}{\Delta_3 \sqrt{(1 - z) \frac{1 - z}{4z} q - \frac{1 - z}{16(y + z)} (x - q)^2}} \int_{q_1}^{q_2} dq.
\]

Since $q_1$ and $q_2$ are roots of equation (4.10), we have

* Use the formula $f_{x+y}(x) = \int f(x-q, q) dq$ where $f_{x+y}(.)$ and $f(.,.)$ are the densities of $x+y$ and $(x,y)$ respectively.

** Note that the integrand is zero outside the interval $[q_1, q_2]$ by the indicator factor.
\[
f_1(x) = \frac{(1-z)^{5/2}}{64\pi z^2/y + z} \frac{\Delta_1 + \Delta_2 q}{\Delta_3 \frac{1-z}{4z} q} q_2^* \frac{q_2 - q_1}{q_1 q_2}.
\]

(4.12)

From (4.10), we can compute
\[
q_1 q_2 = -\frac{x^2}{y + z - yz}
\]
\[
q_2 - q_1 = \frac{2z}{y + z - yz} [(y + z)(-x^2(1 - z) + 4xz + 4(y + z))]^{1/2}
\]
\[
(4.13)
\]
\[
(4.14)
\]

From (4.12) - (4.14), it follows that
\[
f_1(x) = \frac{(1-z)^{5/2}}{64\pi z^2/y + z} \frac{x^2(1-z)/4(y+z)}{(1-z)^2(y+z + xz)} \frac{2z}{16z^2(y+z) x^2} \frac{1-z}{4z} x_2.
\]
\[
= \frac{\sqrt{1-z}}{2\pi(y+z+zx)} \frac{\sqrt{-x^2(1-z) + 4xz + 4(y+z)}}{1-z}.
\]
\[
(4.15)
\]

From this we can easily obtain the density of \[
\left[ 1 + y + \frac{4y^2}{1-z} + \frac{4y y + z}{1-z} \right] / (1-z)
\]
as
\[
f_2(x) = \frac{(1-z)\sqrt{(x-a)(b-x)} x^2 + 2(1+y+z-yz)x - (1-y)^2}{2\pi(xz+y)}
\]
\[
\]
\[
(4.16)
\]

where a = \(\frac{(1-z)\sqrt{y+z-yz})^2}{(1-z)^2}\) and b = \(\frac{(1+y+z-yz)^2}{(1-z)^2}\).

Since \(f_1(x) \neq 0\) iff equation (4.10) has two different roots, we find by checking the steps of computation that \(f_2(x) \neq 0\) iff \(a < x < b\). Recalling (4.9), we obtain
\[
E_k = \frac{1}{1-z} \int_a^b x^{k-1} f_2(x) dx
\]
\[
= \int_a^b x^k \left[ \frac{f_2(x)}{x(1-z)} \right] dx.
\]

(4.17)

* Note that \(\frac{(1-z)}{4z} q - \frac{(1-z)}{16(y+z)} (x-q)^2 = \frac{(1-z)}{4z} q^2\) for \(q = q_1\) or \(q_2\).
Now, let
\[ f(x) = \frac{f_2(x)}{x(1 - z)} = \begin{cases} \frac{(1 - z)^{\sqrt{(x - a)(b - x)}}}{2\pi x(y + xz)} & \text{if } a < x < b, \\ 0 & \text{otherwise.} \end{cases} \tag{4.18} \]

In [7], it is shown that the distribution \( F_{y \cdot z} \) is determined by all its moments.

From (4.11),(4.12) it follows that
\[ F_{y \cdot z}(x) = \gamma I((0,\infty))(x) + \int_{-\infty}^{x} f(x)dx. \tag{4.19} \]

where \( \gamma = 1 - \int_{-\infty}^{\infty} f(x)dx. \)

Finally, we only need to compute the integral \( \int_{a}^{b} f(x)dx. \) If we set
\[ U = \frac{2x - (b + a)}{(b - a)}, \]
we get
\[ I = \int_{a}^{b} f(x)dx = \left( \frac{1 - z}{2\pi y} \right) \left[ \int_{a}^{b} \frac{\sqrt{(x - a)(b - x)}}{x} \ dx - \int_{a}^{b} \frac{\sqrt{(x - a)(b - x)}}{x + \frac{y}{z}} \ dx \right] \]
\[ = \frac{(1 - z)(b - a)}{4\pi y} \left[ \int_{-1}^{1} \sqrt{1 - u^2} \ du - \int_{1}^{b} \frac{\sqrt{1 - u^2}}{u + \lambda_2} \ du \right], \]

where \( \lambda_1 = \frac{b + a}{b - a} \) and \( \lambda_2 = \frac{b + a + (2y)/z}{b - a}. \) Using (2.2) we get
\[ I = \frac{(1 - z)(b - a)}{4y} \left[ (\lambda_1 - \sqrt{\lambda_1^2 - 1}) - (\lambda_2 - \sqrt{\lambda_2^2 - 1}) \right] \]
\[ = \frac{1 - z}{2y} \left[ \sqrt{ab + \frac{Y}{z}(a + b)} + \frac{b^2}{z} - \sqrt{ab - \frac{Y}{z}} \right] \]
\[ = \frac{1 - z}{2y} \left[ \frac{y + z}{z(1 - z)} - \frac{|1 - y|}{1 - z} - \frac{y}{z} \right] \]
\[ = \frac{1}{2y} \left[ 1 + y - |1 - y| \right] \]
\[ = \begin{cases} 1 & \text{if } 0 < y \leq 1 \\ \frac{1}{y} & \text{if } y > 1. \end{cases} \]

Hence
\[ \gamma = \begin{cases} 0 & \\ 1 - \frac{1}{y} \end{cases}. \]

Substituting this into (4.19), we get the expression of \( F_{y \cdot z}(x). \)
ADDED IN PROOF

After this paper was typed, the following interesting paper was brought to the attention of the authors:


In the above paper, Wachter gave an explicit expression for the e.d.f. of $(XJX')(XX')^{-1}$ in the limiting case when $p \to \infty$, $J$ is a projection matrix and the columns of $X$: $p \times (n+m)$ are distributed independently as multivariate normal with mean vector 0 and covariance matrix I. When

$$J = \begin{pmatrix} I_m & 0 \\ 0 & 0 \end{pmatrix},$$

we observe that $(XJX')(XX')^{-1} = (X_1X_1')(X_1X_1' + X_2X_2')^{-1}$ where $X = (X_1 X_2)$

The random matrix considered by us is equivalent to $(X_1X_1')(X_2X_2')^{-1}$ and so our result on the limiting e.d.f. can be obtained from the result of Wachter.

But, the method used by us is entirely different from the method used by Wachter (1980). Also, we derived the moments of the above distribution and these expressions are new.
REFERENCES


