# On the computation of the distribution of the square of the sample multiple correlation coefficient 

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#### Abstract

Two computationally simple methods are proposed to evaluate the distribution function and the density of the square of the sample multiple correlation coefficient. No auxiliary routine is required. The accuracy of recursive computations can be effectively controlled. The distribution function and the density can be evaluated concurrently because their computing formulas are closely related. This property can enhance efficiency of Newton's method for computing the quantiles of the distribution. The corresponding algorithms are provided in a step-by-step form.


Keywords: Central beta distribution; Error bound; Gamma function; Multiple correlation coefficient; Newton's method; Noncentral beta distribution; Quantile; Series representation

## 1. Introduction

Let $X_{1}, X_{2}, \ldots, X_{m}$ have a multivariate normal distribution with mean vector $\mu$ and covariance matrix $\Sigma$, and $R$ be the sample multiple correlation coefficient between $X_{1}$ and $X_{2}, \ldots, X_{m}$ based on a sample of size $N>m$. The density of $Y=R^{2}$ can be expressed as an infinite weighted sum of central beta densities as
follows (see, e.g., Anderson, 1984, p. 145):

$$
\begin{align*}
& f\left(y ; m, N, \rho^{2}\right) \\
& \quad=\sum_{i=0}^{\infty} \frac{\Gamma^{2}[(N-1) / 2+i]\left(\rho^{2}\right)^{i}\left(1-\rho^{2}\right)^{(N-1) / 2} y^{(m-1) / 2+i-1}(1-y)^{(N-m-2) / 2}}{\Gamma[(N-1) / 2] i!\Gamma[(m-1) / 2+i] \Gamma[(N-m) / 2]} \\
& \quad=\sum_{i=0}^{\infty} q_{i} f(y ; a+i, b), \tag{1}
\end{align*}
$$

where $0 \leq y \leq 1, \rho$ is the population multiple correlation coefficient, $\Gamma(\alpha)$ is the gamma function, $a=(m-1) / 2, \quad b=(N-m) / 2, \quad q_{i}=(\Gamma(a+b+i) / \Gamma(a+b) i!)$ $\left(\rho^{2}\right)^{i}\left(1-\rho^{2}\right)^{a+b}$, and $f(y ; a, b)$ is the central beta density with shape parameters $a$ and $b$. The distribution function of $Y$ can, similarly, be expressed as an infinite weighted sum of central beta distribution functions. That is,

$$
\begin{equation*}
P(Y \leq y)=F\left(y ; m, N, \rho^{2}\right)=\sum_{i=0}^{\infty} q_{i} F(y ; a+i, b), \tag{2}
\end{equation*}
$$

where $F(y ; a, b)$ is the central beta distribution function with shape parameters $a$ and $b$. A recursive algorithm for evaluating $F\left(y ; m, N, \rho^{2}\right)$ based on the series representation in (2) was given by Ding and Bargmann (1991a). The algorithm sums up the terms in (2) until the derived upper bound for the error of truncation is less than some predetermined accuracy. Auxiliary routines for evaluating $F(y ; a, b)$ and the natural logarithm of the gamma function are required. Ding and Bargmann (1991b) also computed the quantile $y_{p}$ such that $f\left(y_{p} ; m, N, \rho^{2}\right)=p$ for given values of $m(>1), N(>m), \rho^{2}\left(0 \leq \rho^{2} \leq 1\right)$, and $p(0 \leq p \leq 1)$ by using the Illinois method. In this paper, computationally simple methods are proposed to evaluate $F(y$; $m, N, \rho^{2}$ ) (based on an alternative series representation) and $F\left(y ; m, N, \rho^{2}\right)$ (based on the series representation in (1)). No auxiliary routine is required. The computational accuracy can be effectively controlled by using the error bounds obtained. Moreover, the recurrence formulas for computing $F\left(y ; m, N, \rho^{2}\right)$ and $f\left(y ; m, N, \rho^{2}\right)$ are closely related, and therefore $F\left(y ; m, N, \rho^{2}\right)$ and $f\left(y ; m, N, \rho^{2}\right)$ can be evaluated concurrently. This property can enhance efficiency of Newton's method for computing the quantile $y_{p}$. The corresponding algorithms are provided in a step-by-step form. Numerical methods and the resultant algorithms basically follow those given in Ding (1994) for computing the noncentral beta distribution, which is an infinite weighted sum of central beta distributions with Poisson weights.

## 2. Numerical methods

The numerical methods discussed in this section for evaluating $F\left(y ; m, N, \rho^{2}\right)$ and $f\left(y ; m, N, \rho^{2}\right)$ is for $0<y<1$. No computation is needed for $y=0$ or $y=1$ since $f\left(0 ; m, N, \rho^{2}\right)=f\left(1 ; m, N, \rho^{2}\right)=F\left(0 ; m, N, \rho^{2}\right)=0$, and $F\left(1 ; m, N, \rho^{2}\right)=1$.

A recursive formula for evaluating $F(y ; a+i, b)$ in (2) was given by Ding (1994) as follows:

$$
\begin{equation*}
F(y ; a+i, b)=\sum_{k=i}^{\infty} \frac{(1-y)}{(a+b+k)} f(y ; a+k+1, b)=\sum_{k=i}^{\infty} t_{k}, \quad i=0,1, \ldots, \tag{3}
\end{equation*}
$$

where

$$
\begin{align*}
t_{0} & =\frac{\Gamma(a+b)}{\Gamma(a+1) \Gamma(b)} y^{a}(1-y)^{b} \\
t_{i} & =t_{i-1} y(a+b+i-1) /(a+i), \quad i \geq 1 . \tag{4}
\end{align*}
$$

It follows that $F\left(y ; m, N, \rho^{2}\right)$ can be expressed by a new series in term of central beta densities:

$$
\begin{align*}
F\left(y ; m, N, \rho^{2}\right) & =\sum_{i=0}^{\infty} q_{i}\left(\sum_{k=i}^{\infty} t_{k}\right) \\
& =\sum_{i=0}^{\infty}\left(\sum_{k=0}^{i} q_{k}\right) t_{i} \\
& =\sum_{i=0}^{\infty} v_{i} t_{i} \tag{5}
\end{align*}
$$

where the terms are evaluated recursively by

$$
\begin{align*}
& v_{0}=q_{0}=\left(1-\rho^{2}\right)^{a+b} \\
& v_{i}=v_{i-1}+q_{i}, \quad q_{i}=q_{i-1}(a+b+i-1)\left(\rho^{2}\right) / i, \quad i \geq 1, \\
& t_{i}, i \geq 0, \quad \text { as in (4). } \tag{6}
\end{align*}
$$

Since $m$ and $N$ are both integers, the gamma function in (4) is easy to evaluate in a sense that $\Gamma(\alpha)=(\alpha-1)$ ! if $\alpha$ is an integer, and $\Gamma(\alpha)=(\alpha-1) \ldots \frac{1}{2} \sqrt{\pi}$ if $\alpha$ is a half-integer. $F\left(y ; m, N, \rho^{2}\right)$ can be approximated by the finite sum $\sum_{i=0}^{n-1} v_{i} t_{i}$. Let $E F_{n}$ denote the error of truncation. Using the facts that $\sum_{i=0}^{\infty} q_{i}=1$ (see, e.g., Abramowitz and Stegun, 1965, p. 556), and $\sum_{i=n}^{\infty} t_{i} \leq t_{n-1} y(a+b+n-1)$ / $[(a+n)-(a+b+n) y]$ if $a+n>(a+b+n) y$ (see Ding, 1994), we have, under the same condition,

$$
\begin{equation*}
E F_{n}=\sum_{i=n}^{\infty} v_{i} t_{i} \leq \sum_{i=n}^{\infty} t_{i} \leq t_{n-1} y(a+b+n-1) /[(a+n)-(a+b+n) y] . \tag{7}
\end{equation*}
$$

The error bound given above is a decreasing function of $n$ when $a+n>(a+b+n) y$. To evaluate $F\left(y ; m, N, \rho^{2}\right)$, accumulate the terms in (5), computed recursively through (6), until the error bound is not greater than a specified accuracy $\varepsilon$. In fact, the evaluation of $F\left(y ; m, N, \rho^{2}\right)$ requires no auxiliary routine.

The density $f\left(y ; m, N, \rho^{2}\right)$ of $Y$ expressed in (1) is another series in terms of central beta densities. Its terms can be evaluated recursively, and are related to
those of (5) as follows:

$$
\begin{equation*}
f\left(y ; m, N, \rho^{2}\right)=\sum_{i=0}^{\infty} q_{i} f(y ; a+i, b)=\sum_{i=0}^{\infty} q_{i} s_{i} \tag{8}
\end{equation*}
$$

where

$$
\begin{align*}
& s_{0}=\frac{\Gamma(a+b)}{\Gamma(a)} \Gamma(b) \\
& y^{a-1}(1-y)^{b-1}=a t_{0} / y /(1-y) \\
& s_{i}=s_{i-1} y(a+b+i-1) /(a+i-1)  \tag{9}\\
&=t_{i-1}(a+b+i-1) /(1-y), \quad i \geq 1
\end{align*}
$$

$q_{i}$ and $t_{i}, i \geq 0$, are those in (6).
Let $E f_{n}$ be the truncation error at $i=n$ for the series in (8). Since the sequence $\left\{s_{i}\right\}(i \geq n)$ is decreasing when $a+n>(a+b+n) y$, we have, under the same condition,

$$
\begin{equation*}
E f_{n}=\sum_{i=n}^{\infty} q_{i} s_{i}<\sum_{i=n}^{\infty} q_{i} s_{n}=s_{n}\left(1-\sum_{i=0}^{n-1} q_{i}\right)=s_{n}\left(1-v_{n-1}\right) . \tag{10}
\end{equation*}
$$

Likewise, the above error bound is a decreasing function of $n$, and is used to control the accuracy of the evaluation of $f\left(y ; m, N, \rho^{2}\right)$. Again, the evaluation of $f(y$; $m, N, \rho^{2}$ ) requires no auxiliary routine.

Let $G(y)=F\left(y ; m, N, \rho^{2}\right)-p$. The quantile $y_{p}$ is to be obtained by solving the equation $G(y)=0$. Since $G(0)=-p, G(1)=1-p$, and $G$ is strictly increasing in $[0,1]$, the solution $y_{p}$ of $G(y)=0$ is unique. An efficient root-finding method is Newton's method, which requires the evaluations of both of $F\left(y ; m, N, \rho^{2}\right)$ and $G^{\prime}(y)=f\left(y ; m, N, \rho^{2}\right)$. The process is to repeat computing (see, e.g., Kennedy and Gentle, 1980, pp. 72-73)

$$
\begin{equation*}
y_{j+1}=y_{j}-\frac{\left[F\left(y_{j} ; m, N, \rho^{2}\right)-p\right]}{f\left(y_{j} ; m, N, \rho^{2}\right)}, \quad j=0,1, \ldots \tag{11}
\end{equation*}
$$

until $\left|y_{j+1}-y_{j}\right| \leq \delta y_{j+1}$, where $\delta$ is a specified accuracy. It is obvious that $y_{p}=0$ for $p=0$ and $y_{p}=1$ for $p=1$. No computation is needed for these cases. For $0<p<1$, perform iterations (11) with the starting value $y_{0}=0.5$. For each iteration, $F\left(y_{j} ; m, N, \rho^{2}\right)$ and $f\left(y_{j} ; m, N, \rho^{2}\right)$ can be evaluated concurrently rather than independently because their computing formulas (5) and (8) are closely related. This property can greatly enhance the computational efficiency of Newton's method. To ensure the legality of the iterate $y_{j+1}$, use the same adjustments as in Ding (1994): if $y_{j+1} \leq 0$, replace it by $y_{j} / 2$; if $y_{j+1} \geq 1$, replace it by $\left(y_{j}+1\right) / 2$. Note that the evaluations of $F\left(y_{j} ; m, N, \rho^{2}\right)$ and $f\left(y_{j} ; m, N, \rho^{2}\right)$ should be precise enough so that the accuracy of Newton's solution $y_{p}$ can be ensured. Also, the number of iterations should be controlled.

## 3. Algorithms

According to the formulas discussed in Section 2, we provide three effective algorithms in a step-by-step form for evaluating $F\left(y ; m, N, \rho^{2}\right), f\left(y ; m, N, \rho^{2}\right)$, and the quantile $y_{p}$. The algorithm for computing $y_{p}$ is particularly efficient.

Algorithm A. This algorithm computes the distribution function $F\left(y ; m, N, \rho^{2}\right)$ of $Y=R^{2}$ for given values of $y(0<y<1), m(>1), N(>m)$, and $\rho^{2}\left(0 \leq \rho^{2} \leq 1\right)$.

A1 (Specify the accuracy). Set $E P S \leftarrow \varepsilon$ (some desired accuracy, e.g., $10^{-6}$ ).
A2 (Set the constants). Set $a \leftarrow(m-1) / 2, b \leftarrow(N-m) / 2$.
A3 (Initialize). Set $n \leftarrow 1, t \leftarrow(\Gamma(a+b) / \Gamma(a+1) \Gamma(b)) y^{a}(1-y)^{b}, q \leftarrow\left(1-\rho^{2}\right)^{a+b}$, $v \leftarrow q, C D F \leftarrow v t$.

A4 (Compare $a+n,(a+b+n) y$ ). If $a+n>(a+b+n) y$, go to step A6.
A5 (Update the term and then accumulate. Then increase $n$ by 1). Set $q \leftarrow q(a+b+n-1)\left(\rho^{2}\right) / n, \quad v \leftarrow v+q, \quad t \leftarrow t y(a+b+n-1) /(a+n), \quad C D F \leftarrow$ $C D F+v t, n \leftarrow n+1$, and return to step A4.

A6 (Find the error bound and check for convergence). Set bound $\leftarrow$ $t y(a+b+n-1) /((a+n)-(a+b+n) y)$. If bound $\leq E P S$, terminate the algorithm. ( $C D F$ is the answer.)

A7 (Update the term and then accumulate. Then increase $n$ by 1). Set $q \leftarrow q(a+b+n-1)\left(\rho^{2}\right) / n, \quad v \leftarrow v+q, \quad t \leftarrow t y(a+b+n-1) /(a+n), \quad C D F \leftarrow$ $C D F+v t, n \leftarrow n+1$, and return to step A6.

Algorithm B. This algorithm computes the density $f\left(y ; m, N, \rho^{2}\right)$ of $Y=R^{2}$ for given values of $y(0<y<1), m(>1), N(>m)$, and $\rho^{2}\left(0 \leq \rho^{2} \leq 1\right)$.

B1 (Specify the accuracy). Set $E P S \leftarrow \varepsilon$ (some desired accuracy, e.g., $10^{-6}$ ).
B2 (Set the constants). Set $a \leftarrow(m-1) / 2, b \leftarrow(N-m) / 2$.
$\mathbf{B 3}$ (Initialize). Set $n \leftarrow 1, s \leftarrow(\Gamma(a+b) / \Gamma(a) \Gamma(b)) y^{a-1}(1-y)^{b-1}, q \leftarrow\left(1-\rho^{2}\right)^{a+b}$, $v \leftarrow q, P D F \leftarrow q s$.

B4 (Compare $a+n,(a+b+n) y$ ). If $a+n>(a+b+n) y$, go to step B6.
B5 (Update the term and then accumulate. Then increase $n$ by 1). Set $q \leftarrow q(a+b+n-1)\left(\rho^{2}\right) / n, v \leftarrow v+q, s \leftarrow s y(a+b+n-1) /(a+n-1), P D F \leftarrow$ $P D F+q s, n \leftarrow n+1$, and return to step B4.

B6 (Find the error bound and check for convergence). Set bound $\leftarrow$ $s y(a+b+n-1)(1-v) /(a+n-1)$. If bound $\leq E P S$, terminate the algorithm. ( $P D F$ is the answer.)

B7 (Update the term and then accumulate. Then increase $n$ by 1). Set $q \leftarrow q(a+b+n-1)\left(\rho^{2}\right) / n, v \leftarrow v+q, s \leftarrow s y(a+b+n-1) /(a+n-1), P D F \leftarrow$ $P D F+q s, n \leftarrow n+1$, and return to step B6.

Algorithm C. This algorithm computes the quantile $y_{p}$ of the distribution of $Y=R^{2}$ such that $F\left(y_{p} ; m, N, \rho^{2}\right)=p$ for given values of $m(>1), N(>m), \rho^{2}\left(0 \leq \rho^{2} \leq 1\right)$, and $p(0<p<1)$.

C1 (Specify the accuracy and the maximum number of Newton's iterations allowed). Set $E P S \leftarrow \varepsilon$ (some desired accuracy, e.g., $10^{-6}$, for computing $F\left(y ; m, N, \rho^{2}\right)$
and $f\left(y ; m, N, \rho^{2}\right)$ ), $D E L T A \leftarrow \delta$ (some desired accuracy, e.g., $10^{-4}$, for computing $y_{p}$ ), ITRMAX $\leftarrow N_{\max }$ (an integer, e.g., 10 , for controlling the number of Newton's iterations).
$\mathbf{C 2}$ (Set the constants). Set $\quad a \leftarrow(m-1) / 2, \quad b \leftarrow(N-m) / 2, \quad$ coeff $\leftarrow$ $\Gamma(a+b) / \Gamma(a+1) \Gamma(b), q_{0} \leftarrow\left(1-\rho^{2}\right)^{a+b}$.
$\mathbf{C 3}$ (Loop on $k$, Newton's iteration). First initialize $y$ by setting $y \leftarrow 0.5$. Then perform steps C4-C10 for $k=1,2, \ldots, I T R M A X$. (Steps C4-C10 constitute one iteration.)
$\mathbf{C 4}$ (Initialize within each iteration). Set $n \leftarrow 1, t \leftarrow$ coeff $y^{a}(1-y)^{b}, s \leftarrow a t / y /(1-y)$, $q \leftarrow q_{0}, v \leftarrow q, C D F \leftarrow v t, P D F \leftarrow q s$.

C5 (Compare $a+n,(a+b+n) y)$. If $a+n>(a+b+n) y$, go to step C7.
C6 (Update the term and then accumulate for both of $F\left(y ; m, N, \rho^{2}\right)$ and $f(y$; $\left.m, N, \rho^{2}\right)$. Then increase $n$ by 1). Set $q \leftarrow q(a+b+n-1)\left(\rho^{2}\right) / n, v \leftarrow v+q$, $s \leftarrow t(a+b+n-1) /(1-y), \quad t \leftarrow t y(a+b+n-1) /(a+n), \quad C D F \leftarrow C D F+v t$, $P D F \leftarrow P D F+q s, n \leftarrow n+1$, and return to step C5.

C7 (Find the corresponding error bounds). Set bndcdf $\leftarrow t y(a+b+n-1) /$ $((a+n)-(a+b+n) y), b n d p d f \leftarrow t(a+b+n-1)(1-v) /(1-y)$.

C8 (Check for convergence). If $b n d c d f \leq E P S$ and $b n d p d f \leq E P S$, go to step C10.
C9 (Update the terms and then accumulate for $F\left(y ; m, N, \rho^{2}\right)$ and/or $f(y$; $m, N, \rho^{2}$ ). Then increase $n$ by 1). Set $q \leftarrow q(a+b+n-1)\left(\rho^{2}\right) / n, v \leftarrow v+q$. If $b n d c d f \leq E P S, \quad$ set $\quad s \leftarrow s y(a+b+n-1) /(a+n-1), \quad P D F \leftarrow P D F+q s$, $n \leftarrow n+1$, bndpdf $\leftarrow s y(a+b+n-1)(1-v) /(a+n-1)$, and return to step C8; otherwise if $b n d p d f \leq E P S$, set $t \leftarrow t y(a+b+n-1) /(a+n), C D F \leftarrow C D F+v t$, $n \leftarrow n+1, b n d c d f \leftarrow t y(a+b+n-1) /((a+n)-(a+b+n) y)$, and return to step C8; otherwise set $s \leftarrow t(a+b+n-1) /(1-y), t \leftarrow t y(a+b+n-1) /(a+n)$, $C D F \leftarrow C D F+v t, P D F \leftarrow P D F+q s, n \leftarrow n+1$, and return to step $C 7$.

C10 (Find new $y$ and check for convergence of Newton's process). Set diff $\leftarrow(C D F-p) / P D F$. If $y-$ diff $\leq 0$, set $y \leftarrow y / 2$; otherwise if $y$-diff $\geq 1$, set $y \leftarrow(y+1) / 2$; otherwise set $y \leftarrow y$-diff. If $\mid$ diff $\mid / y \leq D E L T A$, terminate the algorithm. ( $y$ is the answer.)

C11 (Output error message). Terminate the algorithm with the message "No convergence after $N_{\max }$ iterations".

FORTRAN codes based on the above algorithms are available upon request.

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