

Approximations for the quantiles of Student's t and F distributions and their error bounds

Yasunori FUJIKOSHI and Satoru MUKAIHATA
(Received April 22, 1992)

1. Introduction

Let $F_n(x)$ be the distribution function of a random variable X_n depending on some parameter n , not necessary a sample size. A typical form of the asymptotic expansion of $F_n(x)$ around the limiting distribution function $G(x)$ of $F_n(x)$ as $n \rightarrow \infty$ is

$$(1.1) \quad F_n(x) = G(x) + g(x) \left\{ \frac{1}{n} a_1(x) + \frac{1}{n^2} a_2(x) + \cdots \right\},$$

or the one with n replaced by \sqrt{n} , where $g(x)$ is the density function of $G(x)$, and $a_1(x)$, $a_2(x)$, etc. are suitable polynomials. When $F_n(x)$ is approximated by a function of the form

$$G_{k,n}(x) = G(x) + g(x) \sum_{j=1}^{k-1} a_j(x) n^{-j},$$

it is well known that the error $R_{k,n}(x) = F_n(x) - G_{k,n}(x)$ satisfies

$$R_{k,n}(x) = O(n^{-k})$$

under suitable regularity conditions (see, e.g., Bhattacharya and Ghosh [1]). This means that there exists a positive constant C_k such that for large n

$$|R_{k,n}(x)| \leq C_k n^{-k}$$

However, such C_k and n have not been obtained except for some special statistics (see, e.g., Fujikoshi and Shimizu [3]).

On the other hand, it is also important to find out $x_n(u)$ such that

$$(1.2) \quad F_n(x_n(u)) = G(u).$$

Based on the expansion (1.1), we can formally expand $x_n(u)$ as

$$(1.3) \quad x_n(u) = u + \frac{1}{n} b_1(u) + \frac{1}{n^2} b_2(u) + \cdots.$$

This is usually done, first by finding out $u_n(x)$ of the form

$$(1.4) \quad u_n(x) = x + \frac{1}{n} \tilde{a}_1(x) + \frac{1}{n^2} \tilde{a}_2(x) + \cdots$$

which satisfies

$$(1.5) \quad F_n(x) = G(u_n(x)),$$

and then by solving the equation (1.4) of x (see, e.g., Hill and Davis [4]). Expansions (1.3) and (1.4) are called Cornish-Fisher expansions. The quantile $x_n(u)$ of $F_n(x)$ is usually approximated by a function of the form

$$x_{k,n}(u) = u + \sum_{j=1}^{k-1} b_j(u) n^{-j}.$$

However, it is little known about the error estimate of this approximation. An ideal result is to find out an upper bound $\bar{l}_n(u)$ and a lower bound $l_n(u)$ of $x_n(u)$, such that

$$(1.6) \quad l_n(u) \leq x_n(u) \leq \bar{l}_n(u)$$

and

$$(1.7) \quad 0 \leq \bar{l}_n(u) - l_n(u) \leq D_k n^{-k},$$

where D_k is a positive constant. In general, it will be difficult to have an error estimate of the form (1.7). So, as a more feasible form we will consider upper and lower bounds such that

$$(1.8) \quad 0 \leq \bar{l}_n(u) - l_n(u) \leq \frac{D}{n} |u|,$$

where D is a positive constant. In fact, Wallace [5] obtained upper and lower bounds in the forms (1.6) and (1.7) with $k = \frac{1}{2}$ and (1.8) for $u_n(x)$ of Student's t distribution.

It may be noted that Wallace's results are for $u_n(x)$, not for $x_n(u)$ of Student's t distribution. In this paper, we give an approximation, which has an error estimate in the form (1.8), for $x_n(u)$ of t and F distributions. Our approximations will be proposed with the help of Cornish-Fisher expansions. The proofs are based on the method of Wallace [5].

2. Preliminary results

Let F and G be absolutely continuous distribution functions with density functions f and g , respectively, and let $x(u)$ be the solution of the equation $F(x) = G(u)$ for x in terms of u .

Assume throughout that the density $g(u)$ are positive and continuous for

$c < u < \infty$ and that an approximation $z(u)$ to $x(u)$ is a continuously differentiable, strictly increasing function for $c < u < \infty$. Here c is any appropriately chosen constant which can be $-\infty$. Further, assume that the density $f(x)$ is continuous for $\lim_{u \rightarrow c} z(u) < x < \infty$. Let

$$(2.1) \quad R(u) = \frac{f(z(u))z'(u)}{g(u)}.$$

The following theorem and lemma were proved by Wallace [5].

THEOREM 2.1. *If*

- (a₁) $\lim_{u \rightarrow \infty} z(u) = \infty$
- (a₂) $\lim_{u \rightarrow c} G(u) = F(\lim_{u \rightarrow c} z(u))$
- (a₃) $\text{sgn}\{R(u) - 1\}$ is monotonic function of u for $c < u < \infty$,

then $x(u) \geq z(u)$ or $x(u) \leq z(u)$ for all $c < u < \infty$ according as the function in (a₃) is increasing or decreasing.

LEMMA 2.1. *For all $y > 0$, $h_d(y) = (e^y - 1)/(ye^{dy})$ is monotone decreasing for $d \geq 1$, monotone increasing for $0 < d \leq \frac{1}{2}$ and not monotonic for $\frac{1}{2} < d < 1$.*

3. Student's t distribution

Let F_n, f_n be respectively the distribution and the density functions of Student's t with n degrees of freedom, let Φ, ϕ be respectively the standard normal distribution and the density functions, and let $x_n(u)$ be the solution of the equation

$$(3.1) \quad F_n(x) = G(u)$$

for x in terms of u . It is well known that $x_n(u)$ can be formally expanded as

$$x_n(u) = u + \frac{1}{4n}(u^3 + u) + \frac{1}{96n^2}(5u^5 + 16u^3 + 3u) + \dots$$

Let $l_n(u)$ and $\bar{l}_n(u)$ be two approximations to $x_n(u)$ defined by

$$(3.2) \quad \begin{aligned} l_n(u) &= n^{1/2}(e^{u^2/n} - 1)^{1/2} \\ &= u + \frac{1}{4n}u^3 + \frac{5}{96n^2}u^5 + \dots \end{aligned}$$

and

$$(3.3) \quad \begin{aligned} \bar{l}_n(u) &= n^{1/2}\{e^{(1/n)(1-(2n))^{-1}u^2} - 1\}^{1/2} \\ &= u + \frac{1}{4n}(u^3 + u) + \frac{1}{96n^2}(5u^5 + 18u^3 + 9u) + \dots, \end{aligned}$$

respectively. We note that these two approximations can be intuitively proposed by looking the first two terms in an expansion of $x_n(u)$. Then,

THEOREM 3.1. For all $u > 0$,

- (i) $x_n(u) \geq l_n(u) \quad (n > 0)$;
- (ii) $x_n(u) \leq \bar{l}_n(u) \quad (n > \frac{1}{2})$.

PROOF. Let $z_n(u) = n^{1/2}(e^{\lambda u^2/n} - 1)^{1/2}$ with a positive constant λ . Then it is easily seen that $z_n(u)$ is continuously differentiable, strictly increasing for $0 < u < \infty$. We can write $R_n(u) = f_n(z_n(u))z'_n(u)/\phi(u)$ as

$$R_n(u) = \frac{\Gamma((n+1)/2)}{\Gamma(n/2)(n/2)^{1/2}} \left(\frac{e^y - 1}{ye^{dy}} \right)^{-1/2},$$

where $y = \lambda u^2/n$ and $d = 1 - n(1 - \lambda^{-1})$. Note that y is a monotone increasing function of u for all $u > 0$. By Lemma 2.1 we have that $R_n(u)$ is monotone increasing for $d = 1$ or $\lambda = 1$ and decreasing for $d = 1/2$ or $\lambda = (1 - 1/(2n))^{-1}$. Hence inequalities (i), (ii) follows from Theorem 2.1. Q.E.D.

Let X_n be a random variable whose distribution is Student's t with n degrees of freedom. We consider a transformed random variable

$$(3.4) \quad Y_n = \left\{ \left(n - \frac{1}{2} \right) \log \left(1 + \frac{1}{n} X_n^2 \right) \right\}^{1/2} \text{sgn}(X_n), \quad \text{for } n > \frac{1}{2}.$$

This variable has a rapid convergence to the standard normal distribution, in a sense of

$$P(Y_n \leq y) = \Phi(y) + O(n^{-2})$$

for all real y . Let $y_n(u)$ be the solution of the equation

$$P(Y_n \leq y) = \Phi(u)$$

for y in terms of u . Then,

THEOREM 3.2. For all $u > 0$,

- (i) $\left(1 - \frac{1}{2n} \right)^{1/2} u \leq y_n(u) \leq u \quad \left(n > \frac{1}{2} \right)$;
- (ii) $0 \leq u - \left(1 - \frac{1}{2n} \right)^{1/2} u \leq n_0 \left\{ 1 - \left(1 - \frac{1}{2n_0} \right)^{1/2} \right\} \frac{u}{n} \quad \left(n \geq n_0 > \frac{1}{2} \right)$.

PROOF. We have

$$y_n(u) = \left\{ \left(n - \frac{1}{2} \right) \log \left(1 + \frac{1}{n} x_n^2(u) \right) \right\}^{1/2}$$

for all $u > 0$. Therefore, (i) follows immediately from Theorem 3.1. Inequality (ii) follows from

$$1 - \left(1 - \frac{1}{2n}\right)^{1/2} \leq \frac{n_0}{n} \left\{1 - \left(1 - \frac{1}{2n_0}\right)^{1/2}\right\}$$

for all $n \geq n_0 > \frac{1}{2}$ which has been proved in Wallace [5]. Q.E.D.

4. F distribution

Let F_n and f_n be the distribution and the density functions of a random variable $X_n = \chi_q^2 / (\chi_n^2/n)$, respectively, where χ_q^2 and χ_n^2 are mutually independent chi-square variables with q and n degrees of freedom, respectively. Let G and g be the distribution and the density functions of χ_q^2 , respectively. It is well known (see, e.g., Fujikoshi [2]) that

$$(4.1) \quad F_n(x) = G(x) - g(x) \left[\frac{1}{n} \left(\frac{1}{2}x^2 - \frac{1}{2}(q-2)x \right) + \frac{1}{n^2} \left\{ \frac{1}{16}x^4 - \frac{1}{48}(9q-2)x^3 + \frac{1}{48}(q-2)(9q-4)x^2 - \frac{1}{16}(q-2)(q-4)(3q-2)x \right\} + \dots \right].$$

First, let $u_n(x)$ be the solution of the equation

$$(4.2) \quad F_n(x) = G(u)$$

for u in terms of x (i.e., $u_n(x)$ is the chi-square deviate corresponding to the argument x of F_n). Then

$$(4.3) \quad u_n(x) = x - \frac{1}{n} \left(\frac{1}{2}x^2 - \frac{1}{2}(q-2)x \right) + \frac{1}{n^2} \left\{ \frac{1}{3}x^3 - \frac{7}{24}(q-2)x^2 - \frac{1}{24}(q-2)(q+2)x \right\} - \dots.$$

Let $w_{1,n}(x)$ and $w_{2,n}(x)$ be two approximations to $u_n(x)$ defined by

$$(4.4) \quad w_{1,n}(x) = n \log \left(1 + \frac{x}{n} \right) = x - \frac{x^2}{2n} + \frac{x^3}{3n^2} - \dots$$

and

$$(4.5) \quad w_{2,n}(x) = \left\{ n + \frac{1}{2}(q-2) \right\} \log \left(1 + \frac{x}{n} \right) = x - \frac{1}{n} \left(\frac{1}{2}x^2 - \frac{1}{2}(q-2)x \right) + \frac{1}{n^2} \left(\frac{1}{3}x^3 - \frac{1}{4}(q-2)x^2 \right) - \dots,$$

respectively. We note that the right-hand sides of (4.4) and (4.5) are closely related to the one of (4.3). Then

THEOREM 4.1. For all $x > 0$,

- (i) $u_n(x) \leq w_{1,n}(x)$ ($0 < q < 2$ and $n > 0$);
 $u_n(x) \geq w_{2,n}(x)$ ($0 < q < 2$ and $n > (2 - q)/2$);
- (ii) $u_n(x) = w_{1,n}(x) = w_{2,n}(x)$ ($q = 2$ and $n > 0$);
- (iii) $w_{1,n}(x) \leq u_n(x) \leq w_{2,n}(x)$ ($q > 2$ and $n > 0$);
- (iv) $|w_{1,n}(x) - w_{2,n}(x)| \leq \frac{|q - 2| x}{2 n}$ ($q > 0$ and $n > 0$).

PROOF. Let $z_n(x) = \lambda n \cdot \log(1 + x/n)$, where λ is a positive constant to be chosen. Then consider the function $R_n(x) = g(z_n(x))z'_n(x)/f_n(x)$ which can be written as a function of $y = \log(1 + x/n)$ as follows:

$$R_n(x) = C_n \lambda^{q/2} \frac{y^{q/2-1} \exp \left[\left\{ \frac{1}{2}q - 1 + \frac{1}{2}n(1 - \lambda) \right\} y \right]}{(e^y - 1)^{q/2-1}}.$$

where

$$(4.6) \quad C_n = \frac{\Gamma(n/2)(n/2)^{q/2}}{\Gamma((q+n)/2)}$$

First, set $\lambda = 1$. Then $z_n(x) = w_{1,n}(x)$ and

$$R_n(x) = C_n \left(\frac{e^y - 1}{ye^y} \right)^{1-q/2}$$

which is monotone decreasing for $0 < q \leq 2$ and increasing for $q \geq 2$ from Lemma 2.1.

Next set $\lambda = 1 + (q - 2)/(2n)$. Then $z_n(x) = w_{2,n}(x)$ and

$$R_n(x) = C_n \left(1 + \frac{q - 2}{2n} \right)^{q/2} \left(\frac{e^y - 1}{ye^{y/2}} \right)^{1-q/2}$$

which is monotone increasing for $0 < q \leq 2$ and decreasing for $q \geq 2$.

Hence inequalities (i), (ii) and (iii) follow from Theorem 2.1. Finally, (iv) follows immediately from the definitions of $w_{1,n}(x)$ and $w_{2,n}(x)$. Q.E.D.

Next we consider the quantile $x_n(u)$ of F_n , i.e., the solution of the equation (4.2) for x in terms of u . Then

$$(4.7) \quad x_n(u) = u + \frac{1}{n} \left(\frac{1}{2} u^2 - \frac{1}{2} (q-2)u \right) + \frac{1}{n^2} \left\{ \frac{1}{6} u^3 - \frac{11}{24} (q-2)u^2 + \frac{1}{24} (q-2)(7q-10)u \right\} + \dots$$

Let $l_{1,n}(u)$ and $l_{2,n}(u)$ be two approximations to $x_n(u)$ defined by

$$(4.8) \quad l_{1,n}(u) = n(e^{u/n} - 1) = u + \frac{u^2}{2n} + \frac{u^3}{6n^2} + \dots$$

and

$$(4.9) \quad \begin{aligned} l_{2,n}(u) &= n\{e^{1/n(1+(q-2/2n)^{-1}u)} - 1\} \\ &= u + \frac{1}{n} \left(\frac{1}{2} u^2 - \frac{1}{2} (q-2)u \right) \\ &\quad + \frac{1}{n^2} \left\{ \frac{1}{6} u^3 - \frac{1}{2} (q-2)u^2 + \frac{1}{4} (q-2)^2 u \right\} + \dots, \end{aligned}$$

respectively. These approximations may be proposed by comparing the expansion (4.7) with the expansions (4.8) and (4.9). Then,

THEOREM 4.2. For all $u > 0$,

- (i) $x_n(u) \geq l_{1,n}(u)$ ($0 < q < 2$ and $n > 0$);
 $x_n(u) \leq l_{2,n}(u)$ ($0 < q < 2$ and $n > \frac{1}{2}(2-q)$);
- (ii) $x_n(u) = l_{1,n}(u) = l_{2,n}(u)$ ($q = 2$ and $n > 0$);
- (iii) $l_{2,n}(u) \leq x_n(u) \leq l_{1,n}(u)$ ($q > 2$ and $n > 0$).

PROOF. Let $z_n(u) = n(e^{\lambda u/n} - 1)$, where λ is a positive constant to be chosen. Then we can write $R_n(u) = f_n(z_n(u))z'_n(u)/g(u)$ as

$$R_n(u) = C_n^{-1} \lambda^{q/2} \frac{(e^y - 1)^{q/2-1}}{y^{q/2-1} \exp \left[\left\{ \frac{1}{2} q - 1 + \frac{1}{2} n(1 - \lambda^{-1}) \right\} y \right]},$$

where $y = \lambda u/n$ and C_n is defined by (4.6). Hence inequalities (i), (ii) and (iii) follow from Theorem 2.1 and Lemma 2.1 Q.E.D.

Consider a transformed random variable

$$Y_n = \left\{ n + \frac{1}{2}(b-2) \right\} \log \left(1 + \frac{1}{n} X_n \right),$$

where $n > \max\{0, (2-b)/2\}$. We note that this transformation is based on

a Bartlett adjustment for a log-likelihood ratio statistic in a linear model. It is known that

$$P(Y_n \leq y) = G(y) + O(n^{-2})$$

for all real y . Let $y_n(u)$ be the solution of the equation

$$P(Y_n \leq y) = G(u)$$

for y in terms of u . Then,

THEOREM 4.3. For all $u > 0$,

- (i) $\left(1 + \frac{q-2}{2n}\right)u \leq y_n(u) \leq u$ ($0 < q < 2$ and $n > \frac{1}{2}(2-q)$);
- (ii) $y_n(u) = u$ ($q = 2$ and $n > 0$);
- (iii) $u \leq y_n(u) \leq \left(1 + \frac{q-2}{2n}\right)u$ ($q > 2$ and $n > 0$).

PROOF. It holds that

$$y_n(u) = \left\{n + \frac{1}{2}(q-2)\right\} \log\left(1 + \frac{1}{n}x_n(u)\right)$$

for all $u > 0$ and $n > \max\left\{\frac{1}{2}(2-q), 0\right\}$. Hence, inequalities (i), (ii) and (iii) follow immediately from Theorem 4.2. Q.E.D.

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*Department of Mathematics
Faculty of Science
Hiroshima University
Higashi-Hiroshima 724, Japan*