

Computable Error Bounds for Approximations of Transformed Chi-Squared Variables and Its Statistical Applications

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Abstract

We get computable error bound of order $O(n^{-1})$ for chi-squared with 1 degree of freedom approximation of transformed chi-squared random variable with n degrees of freedom. The result is applied for likelihood ratio statistics in multivariate case.

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Abbreviated title: Computable error bound of chi-squared approximation

1 Introduction

Let \mathcal{X}_n^2 be a random variable having chi-squared distribution with n degrees of freedom and density

$$p_{\mathcal{X}_n^2}(x) = \frac{1}{2^{n/2} \Gamma(n/2)} x^{-1+n/2} e^{-x/2} I_{(0, \infty)}(x),$$

where $I_A(x)$ denotes indicator function of set A . We consider a transformed chi-squared statistic defined by

$$T_1 = \mathcal{X}_n^2 - n \log \frac{\mathcal{X}_n^2}{n} - n.$$

The distribution of T_1 appears as the null distribution of LR (likelihood ratio) statistic for testing a hypothesis that the variance σ^2 is equal to a given value, based on a sample of $n + 1$ observations from a normal population $N(\mu, \sigma^2)$. Note that for large n

$$P(T_1 \leq x) = G_1(x) + O(n^{-1}),$$

where $G_m(x)$ is the distribution function of chi-squared random variable \mathcal{X}_m^2 with m degrees of freedom. One of our main purposes is to show that

$$\sup_x |P(T_1 \leq x) - G_1(x)| \leq B(n),$$

where $B(n)$ is a computable constant and $B(n) = O(n^{-1})$ as $n \rightarrow \infty$.

In this paper we also obtain Berry-Esseen type bound for asymptotic approximation of the distribution of

$$T_p = \text{tr}W - n \log \left| \frac{1}{n} W \right| - np,$$

which is an extension of T_1 , where the random matrix W has a Wishart distribution $W_p(n, I_p)$. The distribution of T_p appears as the null distribution of LR statistic for testing a hypothesis that the covariance matrix Σ is equal to a given covariance matrix. In fact, we prove that

$$\sup_x |P(T_p \leq x) - G_q(x)| \leq B(p, n),$$

where $q = \frac{1}{2}p(p+1)$ and $B(p, n)$ is a computable constant, depending only on p and n and $B(p, n) = O(n^{-1})$ as $n \rightarrow \infty$.

In Section 2 we formulate the main results. In Section 3 we give an outline for the method of deriving Theorem 1. It is shown that the result will be obtained by estimating the three integrals J_1 , J_2 and J_3 . In Section 5 we estimate the summands J_1 , J_2 and J_3 , based on auxiliary results given in Section 4. A proof of Theorem 1 is given in Section 6. In Section 7 we show that Theorem 2 can be obtained from Theorem 1 with help of some basic properties. The possible generalizations of Theorem 1 are discussed in Section 8.

2 Main Results

Let $p \geq 1$ and $n \geq 1$ be integers and $p \leq n$. Put

$$D(n) = \frac{1.9}{n} \left(\frac{n}{n-1} \right)^2 + \frac{15.59}{n} 0.9906^n + C_1(n) 0.9894^n$$

with $C_1(n) = 15.21 / (n-4)$ for $n > 32$ or $C_1(n) = 0.5271$ for $4 \leq n \leq 32$.

Theorem 1 *We have*

$$\sup_x |P(T_1 \leq x) - G_1(x)| \leq B(n), \quad (1)$$

where

$$B(n) = 2D(n) + \frac{1.877}{n} + \frac{1.1284}{\sqrt{n}} 0.7788^n.$$

Remark 1. It is easy to see that $B(n) = O(n^{-1})$ as $n \rightarrow \infty$.

Theorem 2 *We have*

$$\sup_x |P(T_p \leq x) - G_q(x)| \leq B(p, n), \quad (2)$$

where

$$B(p, n) = \sum_{i=1}^p B(n_i),$$

Remark 2. It is easy to see that $B(p, n) = O(n^{-1})$.

n	$B(n)$	$B(2, n)$	$B(3, n)$
100	0.28796	0.58123	0.87994
200	0.07057	0.14192	0.21407
300	0.02932	0.05884	0.08857
400	0.01710	0.03428	0.05153

3 Outline of the Proof of Theorem 1

Put $h(y) = \sqrt{2n}y - n \log(1 + \sqrt{\frac{2}{n}}y)$. It is easy to see that

$$T_1 = h(V_n) \quad \text{with} \quad V_n = (\mathcal{X}_n^2 - n)/\sqrt{2n}. \quad (3)$$

Note that V_n is a random variable \mathcal{X}_n^2 standardized with $E(\mathcal{X}_n^2) = n$ and $Var(\mathcal{X}_n^2) = 2n$. Hence, by the central limit theorem the distribution function

$$F_n(x) = P(V_n \leq x) = P(\mathcal{X}_n^2 - n \leq \sqrt{2n}x)$$

tends to the normal law $\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-u^2/2} du$ as $n \rightarrow \infty$.

With $E(V_n^3) = 2\sqrt{2/n}$ and the first-order Chebyshev-Edgeworth expansion

$$\Phi_n(x) = \Phi(x) + \frac{\sqrt{2}(1-x^2)}{3\sqrt{n}} \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$

one finds $F_n(x) = \Phi_n(x) + O(1/n)$ as $n \rightarrow \infty$.

Dobrić and Ghosh (1996) gave a bound of the remainder term (see Example 3 in the mentioned paper with $a = b = 1/2$), proving

$$\sup_x |F_n(x) - \Phi_n(x)| \leq \frac{1.9}{n} \left(\frac{n}{n-1} \right)^2 + \frac{15.59}{n} 0.9906^n + C_1(n) 0.9894^n \quad (4)$$

with $C_1(n) = 15.21 / (n-4)$ for $n > 32$ or $C_1(n) = 0.5271$ for $4 \leq n \leq 32$.

Define

$$p_{V_n}(x) = \frac{d}{dx} P(V_n \leq x), \quad \varphi(x) = \frac{d}{dx} \Phi(x), \quad \varphi_n(x) = \frac{d}{dx} \Phi_n(x), \quad (5)$$

$$B_x = \{y \in \mathbb{R} : h(y) \leq x\} \quad \text{and} \quad A_x = \{y \in \mathbb{R} : |y| \leq \sqrt{x}\}.$$

Then

$$\begin{aligned}
P(T_1 \leq x) - G_1(x) &= P(h(V_n) \leq x) - [\Phi(\sqrt{x}) - \Phi(-\sqrt{x})] \\
&= \int_{B_x} p_{V_n}(y) dy - \int_{A_x} \varphi(y) dy \\
&= J_1 + J_2 + J_3,
\end{aligned} \tag{6}$$

where

$$\begin{aligned}
J_1 &= \int_{B_x} (p_{V_n}(y) - \varphi_n(y)) dy, \\
J_2 &= \int_{B_x} \varphi(y) dy - \int_{A_x} \varphi(y) dy, \\
J_3 &= \int_{B_x} (\varphi_n(y) - \varphi(y)) dy - \int_{A_x} (\varphi_n(y) - \varphi(y)) dy.
\end{aligned}$$

Here we used

$$\int_{A_x} (\varphi_n(y) - \varphi(y)) dy = 0$$

because A_x is a symmetric set and the function $\varphi_n(y) - \varphi(y)$ is odd.

In Section 5 we get bounds for J_1 , J_2 and J_3 of order $O(n^{-1})$ using (4) and the following facts:

- Lebesgue measure of the set $\{A_x \Delta B_x\}$ is of order $O(n^{-1/2})$;
- the function $\varphi(y)$ is even;
- the function $\varphi_n(y) - \varphi(y)$ is odd and is of order $O(n^{-1/2})$.

See the detailed proof of Theorem 1 in Section 6. The possible generalizations of Theorem 1 are discussed in Section 8.

4 Auxiliary Results

Let $f(y) = y - \log(1 + y)$ for $y > -1$.

Lemma 1 *Let t be a real number such that*

$$0 < t < t_0 = -5 + \sqrt{40} = 1.3245. \tag{7}$$

Assume that y_t and \bar{y}_t satisfy the conditions:

$$f(y_t) = f(\bar{y}_t) = t^2/2 \quad \text{with} \quad y_t > 0 \quad \text{and} \quad \bar{y}_t < 0. \tag{8}$$

Then we have

$$t + \frac{t^2}{3} + \frac{t^3}{45} < y_t < t + \frac{t^2}{3} + \frac{t^3}{36} \quad (9)$$

and

$$-t + \frac{t^2}{3} - \frac{t^3}{27} < \bar{y}_t < -t + \frac{t^2}{3} - \frac{t^3}{36}. \quad (10)$$

Remark 1. Note $f(y)$ is decreasing when $y \in (-1, 0)$ and it is increasing when $y > 0$. Since $f(y)$ is continuous function, $\lim_{y \downarrow -1} f(y) = \lim_{y \uparrow +\infty} f(y) = +\infty$ and $f(0) = 0$, the solutions y_t and \bar{y}_t are defined uniquely by the conditions (8).

Remark 2. One can write the solution $\bar{y}_t < 0$ of $f(\bar{y}_t) = t^2/2$ in terms of the real Lambert's W-function. We recall that the real Lambert's W-function is defined to be the function satisfying $W(u)e^{W(u)} = u$, i.e. it is the inverse function of ze^z for $z \geq -1$, i.e. if $ze^z = u$, then $z = W(u)$. Some properties and applications of the real Lambert's W-function are given in Corless, Gonnet, Hare, Jeffrey and Knuth(1996), e.g. series representation holds

$$W(z) = \sum_{n=1}^{\infty} \frac{(-n)^{n-1}}{n!} z^n = z - z^2 + \frac{3}{2}z^3 - \frac{8}{3}z^4 + \frac{125}{24}z^5 + O(z^6), \quad z \rightarrow 0,$$

which absolutely converges for $|z| < 1/e$. In order to find a solution of the equation

$$f(y) := y - \log(1+y) = t^2/2 \quad (11)$$

provided $-1 < y < 0$, we put $z = 1+y$. Then (11) can be written in equivalent form as

$$(-z)e^{-z} = -\exp\{-1 - t^2/2\}.$$

Therefore, using Taylor expansion with $W^{(k)}(0) = (-k)^{k-1}$ for $k = 0, 1, \dots$ we get

$$\begin{aligned} \bar{y}_t &= -W(-\exp\{-1 - t^2/2\}) - 1 \\ &= -t + (1/3)t^2 - (1/36)t^3 - (1/270)t^4 - (1/4320)t^5 + O(t^6), \quad t \rightarrow 0. \end{aligned}$$

Proof. All four inequalities in (9) and (10) can be proved in the same way. At first we prove the left hand-side inequality of (9). Since $f(y)$ is increasing when $y > 0$, the left-hand side inequality of (9) will be proved if we show that

$$f(t + t^2/3 + t^3/45) < f(y_t) = t^2/2, \quad (12)$$

provided t satisfies (7). Put

$$\begin{aligned} \lambda_1(t) &= f(t + t^2/3 + t^3/45) - t^2/2 \\ &= t - t^2/6 + t^3/45 - \ln(1 + t + t^2/3 + t^3/45). \end{aligned}$$

We have $\lambda_1(0) = 0$. Therefore, in order to prove (12) it is sufficient to show that the derivative $\lambda_1'(t)$ is negative when t satisfies (7). It is easy to see that

$$\lambda_1'(t) = \frac{t^3(t^2 + 10t - 15)}{675(1 + t + t^2/3 + t^3/45)}.$$

The quadratic equation $t^2 + 10t - 15 = 0$ has the solutions $-5 \pm \sqrt{40}$. Hence $\lambda_1'(t) < 0$ and also $\lambda_1(t) < 0$ for t satisfying (7).

Note that $\lambda_1(t) < 0$ for $0 < t < t_0^*$, where t_0^* is the solution of

$$t - t^2/6 + t^3/45 - \ln(1 + t + t^2/3 + t^3/45) = 0,$$

where $1.7 < t_0^* < 1.75$. Hence we may enlarge the interval in (7) until t_0^* .

The other inequality in (9) we find in a similar way. Put

$$\lambda_2(t) = f(t + t^2/3 + t^3/36) - t^2/2.$$

We have $\lambda_2(0) = 0$ and

$$\lambda_2'(t) = \frac{t^4(t + 8)}{432(1 + t + t^2/3 + t^3/36)}.$$

Hence the derivative $\lambda_2'(t)$ is positive and $\lambda_2(t) > 0$.

Since $f(y)$ is decreasing when $y \in (-1, 0)$, the inequalities of (10) will be proved if we show that

$$f(-t + t^2/3 - t^3/27) < f(\bar{y}_t) = t^2/2 < f(-t + t^2/3 - t^3/36) \quad (13)$$

provided t satisfies (7). Put

$$\lambda_3(t) = f(-t + t^2/3 - t^3/27) - t^2/2, \quad \lambda_4(t) = f(-t + t^2/3 - t^3/36) - t^2/2.$$

Then $\lambda_3(0) = 0 = \lambda_4(0)$ and we obtain (13) by

$$\begin{aligned} \lambda_3'(t) &= \frac{t^3(t^2 - 6t + 9)}{243(1 - t + t^2/3 - t^3/27)} > 0 \quad \text{and} \\ \lambda_4'(t) &= \frac{t^4(t - 8)}{432(1 - t + t^2/3 - t^3/36)} < 0. \end{aligned}$$

Lemma 1 is proved. Now we consider a function

$$h(y) = \sqrt{2n} y - n \log \left(1 + \sqrt{2/n} y \right).$$

Obviously,

$$h(y) = n f \left(\sqrt{2/n} y \right). \quad (14)$$

Lemma 2 *Let t_0 be the same as in Lemma 1 and x satisfy*

$$0 < \sqrt{2x/n} < t_0 = -5 + \sqrt{40} = 1.3245. \quad (15)$$

Let $y_1(x)$, $y_2(x)$ be such that

$$h(y_1(x)) = h(y_2(x)) = x \quad \text{with} \quad y_1(x) > 0 \quad \text{and} \quad y_2(x) < 0. \quad (16)$$

Then we have

$$\sqrt{x} + \sqrt{\frac{2}{n}} \frac{x}{3} + \frac{2}{n} \frac{x^{3/2}}{45} < y_1(x) < \sqrt{x} + \sqrt{\frac{2}{n}} \frac{x}{3} + \frac{2}{n} \frac{x^{3/2}}{36} \quad (17)$$

$$-\sqrt{x} + \sqrt{\frac{2}{n}} \frac{x}{3} - \frac{2}{n} \frac{x^{3/2}}{27} < y_2(x) < -\sqrt{x} + \sqrt{\frac{2}{n}} \frac{x}{3} - \frac{2}{n} \frac{x^{3/2}}{36}. \quad (18)$$

Proof: Inequalities (17) and (18) follow from (14) and (9), (10) when we take $t = \sqrt{2x/n}$ in Lemma 1 considering

$$y_t = \sqrt{\frac{2}{n}} y_1(x) \quad \text{and} \quad \bar{y}_t = \sqrt{\frac{2}{n}} y_2(x).$$

Lemma 2 is proved.

5 Bounds for J_1 , J_2 and J_3 from (6) under (15)

Recall that $A_x = (-\sqrt{x}, \sqrt{x})$ and $B_x = \{y : h(y) \leq x\}$. The function $h(y)$ is decreasing for $y \in (-\sqrt{n/2}, 0)$, $h(0) = 0$ and $h(y)$ is increasing for $y > 0$ (cp. Remark 2 after formulation of Lemma 1). Therefore, the set B_x is in fact an interval $(y_2(x), y_1(x))$ according to definitions of $y_1(x)$ and $y_2(x)$ in (16). Now we show how to get bounds for

$$J_1 = \int_{B_x} (pV_n(y) - \varphi_n(y)) dy,$$

$$J_2 = \int_{B_x} \varphi(y) dx - \int_{A_x} \varphi(y) dy, \quad (19)$$

$$J_3 = \int_{B_x} (\varphi_n(y) - \varphi(y)) dy - \int_{A_x} (\varphi_n(y) - \varphi(y)) dy. \quad (20)$$

At first we consider J_1 . We get (see (5))

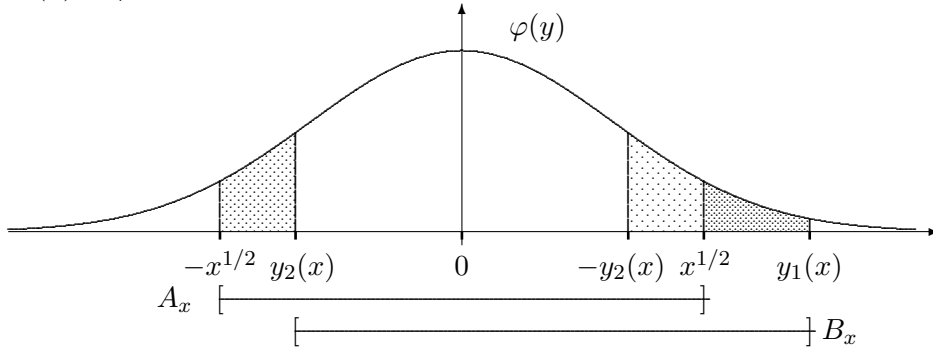
$$J_1 = \left(F_n(y_1(x)) - \Phi_n(y_1(x)) \right) - \left(F_n(y_2(x)) - \Phi_n(y_2(x)) \right),$$

which leads to

$$|J_1| \leq 2 \sup_x |F_n(x) - \Phi_n(x)|. \quad (21)$$

With (4) we obtain the bound for J_1 .

Now we estimate J_2 . It follows from (17) and (18) that $\sqrt{x} - (-y_2(x)) \leq y_1(x) - \sqrt{x}$, therefore J_2 might be either positive or negative.



If $J_2 < 0$ then it follows from Lemma 2 and (19) that

$$\begin{aligned}
|J_2| = -J_2 &= [\Phi(\sqrt{x}) - \Phi(-\sqrt{x})] - [\Phi(y_1(x)) - \Phi(y_2(x))] \\
&= 2\Phi(\sqrt{x}) - [\Phi(y_1(x)) + \Phi(-y_2(x))] \\
&\leq 2\Phi\left(\sqrt{x} + \frac{b_1 + b_2}{2}\right) \\
&\quad - [\Phi(\sqrt{x} + a + b_1) + \Phi(\sqrt{x} - a + b_2)],
\end{aligned}$$

where

$$a = \frac{x}{3} \sqrt{\frac{2}{n}}, \quad b_1 = \frac{x^{3/2}}{45} \frac{2}{n} \quad \text{and} \quad b_2 = \frac{x^{3/2}}{36} \frac{2}{n}.$$

Using the second order Taylor expansion of both functions $\Phi(\sqrt{x} + a + b_1)$ and $\Phi(\sqrt{x} - a + b_2)$ at the point $\sqrt{x} + (b_1 + b_2)/2$ we find

$$|J_2| \leq -\frac{1}{2}(\varphi'(y^*) + \varphi'(y^{**}))\left(a + \frac{b_1 - b_2}{2}\right)^2 \quad (22)$$

with $\sqrt{x} - a + b_2 \leq y^{**} \leq \sqrt{x} + (b_1 + b_2)/2 \leq y^* \leq \sqrt{x} + a + b_1$ and $\varphi'(y) = -\frac{1}{\sqrt{2\pi}} \cdot y \cdot \exp(-y^2/2) < 0$ if $y > 0$.

Note that under condition (15) we have

$$\left(a + \frac{b_1 - b_2}{2}\right)^2 = \frac{2x^2}{9n} \left(1 - \frac{1}{60} \sqrt{\frac{2x}{n}}\right)^2 \leq \frac{2x^2}{9n}.$$

Since $\sqrt{x} \leq y^*$ then we find

$$x^2 (-\varphi'(y^*)) \leq \frac{1}{\sqrt{2\pi}} \max_{y \geq 0} \{y^5 \exp\{-y^2/2\}\} = \frac{1}{\sqrt{2\pi}} \left(\frac{5}{e}\right)^{5/2}.$$

Let us replace condition (15) by stronger one

$$0 < \sqrt{2x/n} \leq 1. \quad (23)$$

We have

$$\sqrt{x} - a + b_2 = \sqrt{x}(1 - u/3 + u^2/36) = \sqrt{x}(1 - u/6)^2 \quad \text{with} \quad u = \sqrt{2x/n} \leq 1.$$

Since $\sqrt{x} - a + b_2 \leq y^{**}$ and

$$\max_{0 < u \leq 1} \frac{(1 - u/60)^2}{(1 - u/6)^8} = \left(\frac{59}{60}\right)^2 \left(\frac{6}{5}\right)^8,$$

we get

$$\begin{aligned}
& - \frac{9n}{2} \varphi'(y^{**}) \left(a + \frac{b_1 - b_2}{2} \right)^2 = - \varphi'(y^{**}) \frac{x^2 (1 - u/6)^8 (1 - u/60)^2}{(1 - u/6)^8} \\
& \leq \frac{1}{\sqrt{2\pi}} \left(\frac{59}{60} \right)^2 \left(\frac{6}{5} \right)^8 \max_{y \geq 0} \{ y^5 \exp\{-y^2/2\} \} \\
& = \frac{1}{\sqrt{2\pi}} \left(\frac{59}{60} \right)^2 \left(\frac{6}{5} \right)^8 \left(\frac{5}{e} \right)^{5/2}.
\end{aligned}$$

By (22) we find

$$|J_2| \leq \frac{1}{9 \cdot \sqrt{2\pi} \cdot n} \left(\frac{5}{e} \right)^{5/2} \left(1 + \left(\frac{59}{60} \right)^2 \left(\frac{6}{5} \right)^8 \right) \leq \frac{1.049}{n}. \quad (24)$$

Now assume that $J_2 > 0$. In this case Lemma 2 implies that

$$\begin{aligned}
0 < J_2 &= \int_{\sqrt{x}}^{y_1(x)} \varphi(y) dy - \int_{-\sqrt{x}}^{y_2(x)} \varphi(y) dy = \int_{\sqrt{x}}^{y_1(x)} \varphi(y) dy - \int_{-y_2(x)}^{\sqrt{x}} \varphi(y) dy \\
&\leq \int_{\sqrt{x}}^{\sqrt{x}+a+b_3} \varphi(y) dy - \int_{\sqrt{x}}^{\sqrt{x}+a-b_4} \varphi(y) dy = \int_{\sqrt{x}+a-b_4}^{\sqrt{x}+a+b_3} \varphi(y) dy,
\end{aligned}$$

where

$$b_3 = \frac{2}{n} \frac{x^{3/2}}{36} \quad \text{and} \quad b_4 = \frac{2}{n} \frac{x^{3/2}}{27} \quad \text{with} \quad b_3 + b_4 = \frac{2}{n} \cdot \frac{7x^{3/2}}{108}.$$

Therefore,

$$J_2 \leq \frac{14}{108n} x^{3/2} \varphi(\sqrt{x}) \leq \frac{14}{108n\sqrt{2\pi}} \left(\frac{3}{e} \right)^{3/2} \leq \frac{0.06}{n}.$$

Comparing this bound with (24) we get that for J_2 the inequality (24) holds when x satisfies (23).

Now we construct the following bound for J_3 :

$$|J_3| \leq \frac{0.828}{n}. \quad (25)$$

Define $m(y) = (y^3 - 3y) e^{-y^2/2}$. Then

$$\varphi_n(y) - \varphi(y) = \frac{1}{\sqrt{2\pi}} \frac{y^3 - 3y}{3\sqrt{n/2}} e^{-y^2/2} = \frac{m(y)}{3\sqrt{n}\pi}.$$

Recall that $y_1(x) > 0$ and $y_2(x) < 0$. Since $m(-y) = -m(y)$, we have

$$J_3 = \int_{\sqrt{x}}^{y_1(x)} \frac{m(y)}{3\sqrt{n}\pi} dy - \int_{-\sqrt{x}}^{y_2(x)} \frac{m(y)}{3\sqrt{n}\pi} dy = \int_{-y_2(x)}^{y_1(x)} \frac{m(y)}{3\sqrt{n}\pi} dy.$$

It follows from (17) and (18) under (23) that

$$y_1(x) \leq \frac{49}{36} \sqrt{x}, \quad -y_2(x) \geq \frac{25}{36} \sqrt{x} \quad \text{and} \quad y_1(x) + y_2(x) \leq \sqrt{\frac{2}{n}} \frac{2x}{3}. \quad (26)$$

Constructing bound for J_3 we consider different cases depending on the value of x . Write

$$J_3 = \sum_{i=1}^5 J_{3i},$$

where $J_{3i} = J_3 \cdot I_{A_i}(x)$ and $I_A(x)$ denotes indicator function of set A . We take

$$\begin{aligned} A_1 &= (11.3001, n/2], & A_2 &= (6.2203, 11.3001], & A_3 &= (5.9877, 6.2203], \\ A_4 &= (3.3834, 5.9877], & A_5 &= (0, 3.3834]. \end{aligned}$$

It is clear that (25) will be proved when we show for $i = 1, \dots, 5$

$$|J_{3i}| \leq \frac{0.828}{n}. \quad (27)$$

The function $m(y)$ has its extreme points at $y = \pm\sqrt{3 \pm \sqrt{6}}$ and we have to consider only the case $y > 0$. Then $m(y) < 0$ if $0 < y < \sqrt{3}$ and $m(y) > 0$ if $y > \sqrt{3}$.

At first we consider J_{31} . If $x \in A_1$ then $\sqrt{3 + \sqrt{6}} \leq (25/36) \sqrt{x}$. Since the function $m(y)$ is decreasing for $y \geq \sqrt{3 + \sqrt{6}}$, we obtain

$$0 < m(y) \leq m(-y_2(x)) \leq m(25\sqrt{x}/36) \quad \text{for} \quad y \in \left(-y_2(x), y_1(x) \right)$$

and

$$\begin{aligned} \int_{-y_2(x)}^{y_1(x)} m(y) dy &\leq m(25\sqrt{x}/36) \left(y_1(x) + y_2(x) \right) \\ &\leq \left(\left(\frac{25}{36} \right)^3 x^{3/2} - 3 \left(\frac{25}{36} \right) x^{1/2} \right) e^{-(25/36)^2 x/2} \sqrt{\frac{2}{n}} \frac{2x}{3}. \end{aligned}$$

The function $v(x) := (a^3 x^{5/2} - 3 a x^{3/2}) e^{-a^2 x/2}$ takes its maximum value at $x^* = (4 + \sqrt{7}) a^{-2}$ and with $a = 25/36$ we find $v(x) \leq v(x^*) = 4.670$.

Hence, in the this case we have

$$|J_{31}| \leq \frac{1}{\sqrt{2\pi}} \frac{4 v(x^*)}{9 n},$$

which leads to (27) for $i = 1$.

Suppose now that $6.2203 < x \leq 11.3001$, i.e. $(25/36)\sqrt{x} > \sqrt{3}$, then with $0 < m(y) \leq m(\sqrt{3 + \sqrt{6}}) = 0.3749$ for $y \in (-y_2(x), y_1(x))$ and $x \leq 11.3001$ we obtain

$$|J_{32}| \leq \frac{1}{\sqrt{2\pi}} \frac{4x}{9n} m(\sqrt{3 + \sqrt{6}}) \leq \frac{0.7512}{n}.$$

Note that $\max_{y>0} |m(y)| = -m(\sqrt{3 - \sqrt{6}}) \leq 1.3802$. Therefore, we get (27) for $i = 5$.

Next we consider J_{34} . Put $x_1 = 3.3834$ and $x_2 = 5.9877$. By (26) we have

$$-y_2(x_1) \geq 1.2736, \quad y_1(x_2) \leq 3.3307, \quad m(3.3307) \leq -m(1.2736) \leq 0.7799$$

and therefore we get (27) for $i = 4$.

Finally, we construct bound for J_{33} . Put $x_3 = 6.2203$. By (26) we have

$$-y_2(x_2) \geq 1.699, \quad y_1(x_3) \leq 3.3947$$

and

$$-m(1.699) \leq m(3.3947) \leq 0.0911,$$

which leads to (27) for $i = 3$, since $|J_{33}| \leq 0.101/n$.

Thus, for all x satisfying (23) we proved (25).

6 Proof of Theorem 1

Note that J_2 and J_3 are estimated in (24) and (25) provided that x satisfies (15), whereas the bound for J_1 is uniform.

Note now the following fact: let $F(x)$ and $G(x)$ be distribution functions and suppose that for some $x_0 > 0$ we have

$$\sup_{|x| \leq x_0} |F(x) - G(x)| \leq \delta \tag{28}$$

and

$$\max\{G(-x_0), 1 - G(x_0)\} \leq \varepsilon. \quad (29)$$

Then

$$\sup_{x \in \mathbb{R}^1} |F(x) - G(x)| \leq \delta + \varepsilon. \quad (30)$$

In fact, (30) follows immediately from (28) and (29) because (29) implies that

$$\max\{F(-x_0), 1 - F(x_0)\} \leq \delta + \varepsilon$$

and

$$\sup_{|x| \geq x_0} |F(x) - G(x)| \leq \max\{G(-x_0), F(-x_0), 1 - F(x_0), 1 - G(x_0)\}.$$

Since $G_1(x) = P(|Y|^2 \leq x)$, and

$$\begin{aligned} P(|Y|^2 > x) &= \frac{2}{\sqrt{2\pi}} \int_{\sqrt{x}}^{\infty} \exp\{-y^2/2\} dy \leq \sqrt{\frac{2}{\pi x}} \int_{\sqrt{x}}^{\infty} y \exp\{-y^2/2\} dy \\ &= \sqrt{\frac{2}{\pi x}} \int_{x/2}^{\infty} \exp(-z) dz = \sqrt{\frac{2}{\pi x}} \exp\{-x/2\} \end{aligned}$$

we get for $x \geq n/2$

$$P(|Y|^2 > x) \leq \sqrt{\frac{4}{\pi n}} \exp\left(-\frac{n}{4}\right).$$

Therefore, by (4), (6), (21), (24), (25), and (30) we obtain the desired bound (1).

7 Proof of Theorem 2

We show how (2) can be obtained from (1). For $i = 1, \dots, p$, put

$$n_i = n - i + 1 \quad \text{and} \quad X_i = \mathcal{X}_{n_i}^2 - n_i \log \frac{\mathcal{X}_{n_i}^2}{n_i} - n_i.$$

Here all the $\mathcal{X}_{n_i}^2$ -variates are independent. Using a well known Bartlett decomposition theorem, see M. Siotani, T. Hayakawa and Y. Fujikoshi (1985) we can write T_p as

$$T_p = \sum_{i=1}^p (X_i + Z_i), \quad (31)$$

where $Z_i \sim \mathcal{X}_{i-1}^2$, $\mathcal{X}_0^2 = 0$ and all Z_i 's and X_i 's are independent.

Now we show that if D_i is such that

$$\sup_x |P(X_i \leq x) - P(U_i \leq x)| \leq D_i \quad (32)$$

with U_i distributed as \mathcal{X}_1^2 then

$$\sup_x |P(T_p \leq x) - G_q(x)| \leq D_1 + \dots + D_p \quad (33)$$

with $q = p(p+1)/2$.

In fact, (33) follows from (31), Lemma 3 (see below) and the fact that the sum of two independent random variables distributed as \mathcal{X}_m^2 and \mathcal{X}_n^2 resp. has chi-square distribution with $m+n$ degrees of freedom.

Lemma 3 *Let X_1, X_2, U_1, U_2 and Z be independent random variables. Let D_1 and D_2 be such that (32) holds for $i = 1, 2$. Then*

$$\sup_x |P(X_1 + X_2 + Z \leq x) - P(U_1 + U_2 + Z \leq x)| \leq D_1 + D_2. \quad (34)$$

Remark 1. We do not make in Lemma 3 any assumptions about a form of distributions of X_1, X_2, U_1, U_2 and Z . Its independence is important only.

Proof(cp. the beginning of the proof of Theorem 3.1 in V. V. Ulyanov, H. Wakaki, Y. Fujihoshi (2005)). Write

$$\begin{aligned} & \sup_x |P(X_1 + X_2 + Z \leq x) - P(U_1 + U_2 + Z \leq x)| \\ & \leq \sup_x |P(X_1 + X_2 + Z \leq x) - P(U_1 + X_2 + Z \leq x)| \\ & \quad + \sup_x |P(U_1 + X_2 + Z \leq x) - P(U_1 + U_2 + Z \leq x)|. \end{aligned} \quad (35)$$

Since for any independent random variables X, U and Z we have

$$\begin{aligned} & \sup_x |P(X + Z \leq x) - P(U + Z \leq x)| \\ & \leq \sup_x E |P(X \leq x - Z | Z) - P(U \leq x - Z | Z)| \\ & \leq \sup_x |P(X \leq x) - P(U \leq x)|, \end{aligned}$$

we get (34) from (35) and Lemma's assumptions.

8 Generalization of Theorem 1

In Theorem 1 we constructed error bound for distribution of T_1 which allowed (see (3)) representation $T_1 = h(V_n)$. Therefore, the possible generalizations can be made in the directions when we replace either function h or random variable V_n or h and V_n simultaneously by similar objects. Here we give generalization connected with replacing of V_n . Concerning V_n we used in the proof of Theorem 1 only the fact that the distribution of V_n can be approximated by the first-order Chebyshev-Edgeworth expansion and error bound of the approximation is known (see (4)). Therefore, the following generalization holds.

Theorem 3 *Let a random variable W_n allow an approximation*

$$\sup_x |P(W_n \leq x) - \Phi_{1n}(x)| \leq B_1(n), \quad (36)$$

where

$$\Phi_{1n}(x) = \Phi(x) + p(EW_n^3, x) \varphi(x) / \sqrt{n},$$

$p(EW_n^3, x)$ is a polynomial depending on the third moment of W_n and moreover $p(EW_n^3, x)$ is an even function and $B_1(n) = O(n^{-1})$ as $n \rightarrow +\infty$. Let $h(y) = \sqrt{2n} y - n \log(1 + \sqrt{\frac{2}{n}} y)$ and $T = h(W_n)$. Then

$$\sup_x |P(T \leq x) - G_1(x)| \leq 2B_1(n) + \frac{c}{n} + \frac{1.1284}{\sqrt{n}} 0.7788^n,$$

where c is a bounded and computable constant depending on the coefficients of the polynomial p .

Remark. Different examples of W_n when W_n is a normalized sum of independent identically distributed random variables and W_n satisfies (36) could be found e.g. in Dobrić and Ghosh (1996). We have noted that Theorem 1 can be applied to LR statistic for testing a hypothesis that the variance σ^2 is equal to a given value in a normal population $N(\mu, \sigma^2)$ in both cases when μ is known and μ is unknown parameter. By using Theorem 3

it is possible to obtain an error bound for the same statistic in a nonnormal population with known μ .

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