

TAIL PROBABILITIES OF NONCENTRAL QUADRATIC FORMS¹

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Let $S(b) = \sum_r \sigma_r^2 \chi_r^2(n_r, b_r^2)$ be a positive linear combination of independent noncentral chi-square random variables. This note derives two representations for the tail probabilities $P[S(b) > x]$, a Taylor series in the noncentrality parameters and a limiting form of this series for large x . An application of the latter result to statistical tests of Cramér-von Mises type is discussed.

1. Introduction. Let $S(b) = \sum_r \sigma_r^2 \chi_r^2(n_r, b_r^2)$, where $b = \{b_r^2; r \geq 1\}$, $1 = \sigma_1^2 > \sigma_2^2 > \dots > 0$, $\sum_r n_r \sigma_r^2 < \infty$, $\sum_r b_r^2 \sigma_r^2 < \infty$, and the $\{\chi_r^2(n_r, b_r^2); r \geq 1\}$ are independent noncentral chi-square random variables whose degrees of freedom and noncentrality parameters are indicated by the arguments n_r, b_r^2 respectively. It is assumed that the range of the summation defining $S(b)$ is either the set of all positive integers or a finite subset of these. This note derives two representations for the tail probabilities $P[S(b) > x]$, a Taylor series in the noncentrality parameters and a limiting form of this series for large x . An application of the latter result to statistical test of Cramér-von Mises type is discussed in Section 3.

2. The representations. Let $G(x) = P[S(0) > x]$ and, more generally, let

$$G_{r_1, r_2, \dots, r_k}(x) P[\sum_r \sigma_r^2 \chi_r^2(n_r + 2 \sum_{j=1}^k \delta(r, r_j)) > x],$$

where $\delta(i, j)$ is the Kronecker delta and the $\{\chi_r^2(\cdot) r \geq 1\}$ are independent central chi-square random variables whose degrees of freedom are given by the argument. Let $G_{r_1, r_2, \dots, r_k}^{(k)}(x)$ denote the k th derivative of $G_{r_1, r_2, \dots, r_k}(x)$ with respect to x .

THEOREM 1. *If $\sum_r b_r^2 < \infty$,*

$$(2.1) \quad P[S(b) > x] = \sum_{k=0}^{\infty} (-1)^k (k!)^{-1} \sum_{r_1, r_2, \dots, r_k} a_{r_1} a_{r_2} \dots a_{r_k} G_{r_1, r_2, \dots, r_k}^{(k)}(x),$$

where $a_r = b_r^2 \sigma_r^2$ and the series converges uniformly in x and uniformly over every set of the form $\{b: \sum_r b_r^2 \leq c\}$. Moreover, for $1 \leq j \leq k$ and $k \geq 1$,

$$(2.2) \quad -2\sigma_{r_j}^2 G_{r_1, r_2, \dots, r_k}^{(k)}(x) = G_{r_1, r_2, \dots, r_k}^{(k-1)}(x) - G_{r_1, \dots, r_{j-1}, r_{j+1}, \dots, r_k}^{(k-1)}(x).$$

PROOF. The characteristic function of $S(b)$ is

$$(2.3) \quad \varphi(t, b) = [\prod_r (1 - 2\sigma_r^2 it)^{-n_r/2}] \exp[\sum_r b_r^2 \sigma_r^2 it(1 - 2\sigma_r^2 it)^{-1}].$$

Since

$$(2.4) \quad P[S(b) > x] - 1 = (2\pi)^{-1} \int_{-\infty}^{\infty} (it)^{-1} (e^{-itx} - 1) \varphi(t, b) dt,$$

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existence of the derivatives $\{G_{r_1, r_2, \dots, r_k}^{(k)}(x); k \geq 1\}$ follows from the inequality $|\varphi(t, 0)| \leq (1 + 4t^2)^{-1}$ and dominated convergence. The relations (2.2) are proved by considering the corresponding characteristic functions. Expanding the exponential factor in (2.3) in powers of the exponent yields a series representation for $\varphi(t, b)$. Let $\varphi_m(t, b)$ denote the sum of the first m terms of this series. A careful analysis establishes $\lim_{m \rightarrow \infty} \int_{-\infty}^{\infty} |t|^{-1} |\varphi(t, b) - \varphi_m(t, b)| dt = 0$ uniformly over every set of the form $\{b: \sum_r b_r^2 \leq c\}$. The series representation (2.1) is implied by this limit and (2.4).

Another representation for $P[S(b) > x]$ obtainable from the characteristic function $\varphi(t, b)$ under the assumption $\sum_r b_r^2 < \infty$ is

$$P[S(b) > x] = \exp[-2^{-1} \sum_r b_r^2] \times \sum_{k=0}^{\infty} (2^k k!)^{-1} \sum_{r_1, r_2, \dots, r_k} b_{r_1}^2 b_{r_2}^2 \dots b_{r_k}^2 G_{r_1, r_2, \dots, r_k}(x),$$

the series converging uniformly in x and uniformly over every set of the form $\{b: \sum_r b_r^2 \leq c\}$. Expanding the exponential factor in this expression into a series, collecting terms and applying (2.2) to the coefficients of these terms, yields (2.1) formally.

If the $\{\sigma_r^2\}$ and $\{n_r\}$ are such that $G(x)$ can be evaluated explicitly, then it is possible, in principle, to calculate the $\{G_{r_1, r_2, \dots, r_k}(x)\}$ by repeated convolution with suitably scaled chi-square distributions having 2 degrees of freedom. Another possibility, when x is large, is to approximate the individual terms of (2.1) by simpler expressions. Let $A(0) = \prod_{r \geq 2} (1 - \sigma_r^2)^{-n_r/2}$. Zolotarev (1961) and Hoeffding (1964) have shown that for large x ,

$$(2.5) \quad G_{r_1, r_2, \dots, r_k}(x) \sim A(0) \prod_{j=1}^k (1 - \sigma_{r_j}^2)^{-(1-\delta(1, r_j))} \times P[\chi^2(n_1 + 2 \sum_{j=1}^k \delta(1, r_j)) > x].$$

By l'Hospital's rule, the same asymptotic relation holds between the k th derivative with respect to x of both sides. Substituting these asymptotic approximations into (2.1), then collecting the terms associated with the various powers of b_1^2 , and finally simplifying the collected coefficients by using

$$(2.6) \quad P[\chi^2(m) > x] \sim (-2)^k \frac{d^k P[\chi^2(m) > x]}{dx^k}; \quad k \geq 1$$

leads heuristically to the following theorem.

THEOREM 2.

$$(2.7) \quad \lim_{x \rightarrow \infty} \frac{P[S(b) > x]}{P[\chi^2(n_1, b_1^2) > x]} = A(b),$$

where $A(b) = A(0) \exp[\sum_{r \geq 2} b_r^2 \sigma_r^2 (2(1 - \sigma_r^2))^{-1}]$ and the convergence is uniform over every set of the form $\{b: \sum_r b_r^2 \sigma_r^2 \leq c\}$.

PROOF. Instead of pursuing the heuristic derivation, we give a direct proof using the general method developed by Hoeffding (1964) in treating $P[S(0) > x]$.

The density of $\chi_1^2(n_1, b_1^2)$ is

$$(2.8) \quad p(x; n_1, b_1^2) = \begin{cases} \exp(-b_1^2/2)2^{-n_1/2}H_{n_1}(b_1 x^{\frac{1}{2}})x^{n_1/2-1} \exp(-x/2) & \text{if } x > 0, \\ = 0, & \text{if } x \leq 0, \end{cases}$$

where

$$(2.9) \quad \begin{aligned} H_{n_1}(b_1 x^{\frac{1}{2}}) &= \sum_{k=0}^{\infty} \frac{(b_1 x^{\frac{1}{2}}/2)^{2k}}{k!(n_1/2 - 1 + k)!} \\ &= (b_1 x^{\frac{1}{2}}/2)^{-n_1/2+1} I_{n_1/2-1}(b_1 x^{\frac{1}{2}}), \end{aligned}$$

b_1 being the positive square root of b_1^2 and $I_\nu(\cdot)$ being the modified Bessel function of the first kind and order ν (see Whittaker and Watson (1927), pages 372–373). Thus

$$(2.10) \quad \begin{aligned} H_{n_1}(b_1 x^{\frac{1}{2}}) &= [\pi^{\frac{1}{2}}\Gamma((n_1 - 1)/2)]^{-1} \\ &\quad \int_0^\pi \cosh [b_1 x^{\frac{1}{2}} \cos(t)] \sin^{n_1-2}(t) dt & \text{if } n_1 \geq 2 \\ &= \pi^{-\frac{1}{2}} \cosh (b_1 x^{\frac{1}{2}}) & \text{if } n_1 = 1. \end{aligned}$$

Let p_U and $p_{S(b)}$ denote the densities of $U = \sum_{r \geq 2} \sigma_r^2 \chi_r^2(n_r, b_r^2)$ and of $S(b)$ respectively. Note that $A(b) = E[\exp(U/2)]$. Since $p_{S(b)}(x) = \int_0^x p(x-y; n_1, b_1^2)p_U(y) dy$, it follows from (2.8) that

$$(2.11) \quad \frac{p_{S(b)}(x)}{p(x; n_1, b_1^2)} = \int_0^x (1 - y/x)^{n_1/2-1} W(x, y) \exp(y/2) p_U(y) dy, \quad x > 0,$$

where $W(x, y) = H_{n_1}(b_1(x-y)^{\frac{1}{2}})/H_{n_1}(b_1 x^{\frac{1}{2}})$ has as its range the interval $[0, 1]$. Thus, if $n_1 \geq 2$, $p_{S(b)}(x) \leq A(b)p(x; n_1, b_1^2)$ for every x , implying $P[S(b) > x] \leq A(b)P[\chi^2(n_1, b_1^2) > x]$.

Choose arbitrary $\delta \in (0, 1)$. The difference $p_{S(b)}(x)/p(x; n_1, b_1^2) - A(b)$, $x > 0$, can be expressed as the sum of four integrals:

$$(2.12) \quad \begin{aligned} V_1 &= \int_0^{\delta x} [(1 - y/x)^{n_1/2-1} - 1] W(x, y) \exp(y/2) p_U(y) dy \\ V_2 &= \int_{\delta x}^x (1 - y/x)^{n_1/2-1} W(x, y) \exp(y/2) p_U(y) dy \\ V_3 &= \int_0^{\delta x} [W(x, y) - 1] \exp(y/2) p_U(y) dy \\ V_4 &= -\int_{\delta x}^{\infty} \exp(y/2) p_U(y) dy. \end{aligned}$$

Now, $|V_1| \leq x^{-1} \max [1, (1 - \delta)^{n_1/2-2}] \cdot |n_1/2 - 1| \cdot E[U \exp(U/2)]$ and $|V_4| \leq x^{-1} \delta^{-1} E[U \exp(U/2)]$, where

$$(2.13) \quad E[U \exp(U/2)] = A(b) [\sum_{r \geq 2} n_r \sigma_r^2 (1 - \sigma_r^2)^{-1} + \sum_{r \geq 2} b_r^2 \sigma_r^2 (1 - \sigma_r^2)^{-2}].$$

For $x > 0$, $0 \leq y/x \leq \delta$ and arbitrary $\beta > 0$,

$$(2.14) \quad 0 \leq 1 - \frac{\cosh(\beta(x-y)^{\frac{1}{2}})}{\cosh(\beta x^{\frac{1}{2}})} \leq x^{-\frac{1}{2}} \beta 2^{-1} (1 - \delta)^{-\frac{1}{2}} y.$$

In view of this and (2.10),

$$|V_3| \leq x^{-\frac{1}{2}} b_1 2^{-1} (1 - \delta)^{-\frac{1}{2}} E[U \exp(U/2)].$$

The remaining integral V_2 satisfies

$$|V_2| \leq x^{-1} \delta^{-2} B(1, n_1/2) \sup_{y \geq \delta x} [y^2 \exp(y/2) p_U(y)]$$

where $B(\cdot, \cdot)$ denotes the usual beta function.

If $U = \sigma_2^2 \chi_2^2(n_2, b_2^2)$, then from (2.8) and (2.10), $|V_2|$ is $o(x^{-1})$, uniformly over every set of the form $\{b : \sum_r b_r^2 \sigma_r^2 \leq c\}$. In general, let p_W denote the density of $W = \sigma_2^2 [\chi_2^2(n_2, b_2^2) + \chi_3^2(n_3, b_3^2)] + \sum_{r \geq 4} \sigma_r^2 \chi_r^2(n_r, b_r^2)$. The densities of $\sigma_3^2 \chi_3^2(n_3, b_3^2)$ and $\sigma_2^2 \chi_3^2(n_3, b_3^2)$ are, respectively, $\sigma_3^{-2} p(\sigma_3^2 x; n_3, b_3^2)$ and $\sigma_2^{-2} p(\sigma_2^{-2} x; n_3, b_3^2)$. Because of (2.8) and (2.10), for $x > 0$,

$$(2.15) \quad \frac{\sigma_3^{-2} p(\sigma_3^{-2} x; n_3, b_3^2)}{\sigma_2^{-2} p(\sigma_2^{-2} x; n_3, b_3^2)} \leq (\sigma_2/\sigma_3)^{n_1} \exp[b_1(\sigma_3^{-1} - \sigma_2^{-1})x^{\frac{1}{2}} - 2^{-1}(\sigma_3^{-2} - \sigma_2^{-2})x],$$

the right side of which is bounded from above for every $x > 0$ by $K(b_1^2) = (\sigma_2/\sigma_3)^{n_1} \exp[b_1^2(\sigma_2 - \sigma_3)2^{-1}(\sigma_2 + \sigma_3)^{-1}]$, a monotone increasing function of b_1^2 . Hence $p_U(x) \leq K(b_1^2)p_W(x)$ for every $x > 0$. Since $n_2 + n_3 \geq 2$, the remark following (2.11) implies in the present context that

$$p_W(x) \leq B(b) \sigma_2^{-2} p(\sigma_2^{-2} x; n_2 + n_3, b_2^2 + b_3^2),$$

where

$$B(b) = [\prod_{r \geq 4} (1 - \sigma_r^2/\sigma_2^2)^{-n_r/2}] \exp[\sum_{r \geq 4} b_r^2 \sigma_r^2 2(\sigma_2^2 - \sigma_r^2)^{-1}].$$

Hence $|V_2|$ is $o(x^{-1})$ uniformly over every set of the form $\{b : \sum_r b_r^2 \sigma_r^2 \leq c\}$.

From the foregoing analysis of the $\{V_i\}$, we conclude that $|p_{S(b)}(x)/p(x; n_1, b_1^2) - A(b)|$ is $O(x^{-\frac{1}{2}})$ uniformly over every set of the form $\{b : \sum_r b_r^2 \sigma_r^2 \leq c\}$. Theorem 2 follows from this fact and Cauchy's mean value theorem.

3. Statistical application. The large sample theory of tests such as the Cramér-von Mises goodness-of-fit and two-sample tests exemplifies the following general situation (see Durbin and Knott (1972) or Beran (1975) for examples). For every integer $v \geq 1$, S_v is a test statistic used to discriminate between hypothesis H_v and alternative K_v , large values of S_v favoring K_v . As $v \rightarrow \infty$, the limiting distributions of S_v under H_v and under K_v are the same as the distributions of $S(0)$ and $S(b)$ respectively, with $b = \{b_r^2; r \geq 1\}$ determined by the sequence of alternatives $\{K_v; v \geq 1\}$. Moreover, S_v can be represented (using Parseval's theorem in the Cramér-von Mises case) as a sum $S_v = \sum_r \sigma_r^2 T_v(r)$ such that under H_v and under K_v , the sequence $\{\sigma_r^2 T_v(r); r \geq 1\}$ converges in distribution in the l_1 -topology to, respectively, the sequence $\{\sigma_r^2 \chi_r^2(n_r); r \geq 1\}$ and the sequence $\{\sigma_r^2 \chi_r^2(n_r, b_r^2); r \geq 1\}$.

Theorem 2 yields some interesting information concerning the asymptotic power of tests based upon S_v . Let $c(\alpha)$, $d_r(\alpha)$ be constants such that $P[S(0) > c(\alpha)] = P[\chi_r^2(n_r) > d_r(\alpha)] = \alpha$ for $0 < \alpha < 1$. Under K_v , the asymptotic powers of the level α S_v -test and the level α $T_v(r)$ -test are, respectively, $P[S(b) > c(\alpha)]$ and $P[\chi_r^2(n_r, b_r^2) > d_r(\alpha)]$.

COROLLARY.

$$(3.1) \quad \lim_{\alpha \rightarrow 0} \frac{P[S(b) > c(\alpha)]}{P[\chi_1^2(n_1, b_1^2) > d_1(\alpha)]} = C(b),$$

where $C(b) = \exp[\sum_{r \geq 2} b_r^2 \sigma_r^2 (2(1 - \sigma_r^2))^{-1}]$ and the convergence is uniform over every set of the form $\{b: \sum_r b_r^2 \sigma_r^2 \leq c\}$.

PROOF. In view of Theorem 2, it is enough to show that

$$(3.2) \quad \lim_{\alpha \rightarrow 0} \frac{P[\chi_1^2[n_1, b_1^2] > d_1(\alpha)]}{A(0)P[\chi_1^2(n_1, b_1^2) > c(\alpha)]} = 1$$

uniformly over every set of the form $\{b_1^2 \leq c\}$. Let $D(\alpha) = c(\alpha) - d_1(\alpha)$. From Theorem 2 specialized to the case $b = 0$, we have, for small α , $P[\chi_1^2(n_1) > d_1(\alpha)] = \alpha \sim A(0)P[\chi_1^2(n_1) > c(\alpha)]$. Since $P[\chi_1^2(n_1) > x] \sim 2p(x; n_1, 0)$ for large x , it follows that $\lim_{\alpha \rightarrow 0} D(\alpha) = 2 \log A(0)$. (For further details regarding this point, see the proof of Theorem 4 in Beran (1975).)

L'Hospital's rule and (2.8) imply

$$(3.3) \quad P[\chi_1^2(n_1, b_1^2) > x] \sim \exp(-b_1^2/2) \Gamma(n_1/2) H_{n_1}(b_1 x^{1/2}) P[\chi_1^2(n_1) > x]$$

for large x . For arbitrary $\beta > 0$,

$$(3.4) \quad 0 \leq 1 - \frac{\cosh(\beta(c(\alpha) - D(\alpha))^{1/2})}{\cosh(\beta(c(\alpha))^{1/2})} \leq 2^{-1} d_1^{-1}(\alpha) D(\alpha) \beta.$$

The desired uniform limit (3.2) follows from (3.3), (3.4), (2.10) and the previous paragraph. This completes the proof.

The corollary above justifies the following general conclusions for sufficiently small α :

(i) The asymptotic power of the level α S_v -test approximately equals or exceeds that of the level α $T_v(1)$ -test, depending upon the values of $\{b_r^2; r \geq 2\}$.

(ii) If $\{K_v\}$ is such that $b_1^2 = 0$ but $b_r^2 \neq 0$ for some $r \geq 2$, the asymptotic power of the level α S_v -test is a small fraction of that of the level α $T_v(r)$ -test.

Thus, although the S_v -test is asymptotically strictly unbiased against every alternative sequence $\{K_v\}$ for which some $b_r^2 \neq 0$, the test is not very efficient when $b_1^2 = 0$ and α is small. Numerical studies by Durbin and Knott (1972) and by Stephens (1973) of the asymptotic powers of the Cramér-von Mises, Anderson Darling, and Watson goodness-of-fit tests under normal location and scale alternatives support these conclusion for $\alpha = .05$.

REFERENCES

- [1] BERAN, R. (1975). Local asymptotic power of quadratic rank tests for trend. *Ann. Statist.* **3** 401-412.
- [2] DURBIN, J. and KNOTT, M. (1972). Components of Cramér-von Mises statistics I. *J. Roy. Statist. Soc. Ser. B* **34** 290-307.
- [3] HOEFFDING, W. (1964). On a theorem of V. M. Zolotarev. *Theor. Probability Appl.* **9** 89-92.
- [4] STEPHENS, M. A. (1973). Components of goodness-of-fit statistics. Unpublished.

- [5] WHITTAKER, E. T. and WATSON, G. N. (1927). *A Course of Modern Analysis*. Cambridge Univ. Press.
- [6] ZOLOTAREV, V. M. (1961). Concerning a certain probability problem. *Theor. Probability Appl.* **6** 201-204.

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