

CALCULATION OF THE DISTRIBUTION FUNCTION  
OF INFINITE QUADRATIC FORMS IN NORMAL VARIABLES

by

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Abstract

Simple but effective methods of computing the distribution function random variables of the form  $\sum_{n=1}^{\infty} \lambda_n Z_n^2$  ( $Z_n$  i.i.d.  $N(0,1)$ ) are described. A FORTRAN subroutine to implement the methods is appended. Some examples are given.

Key Words and Phrases: infinite quadratic form in normal variables,  
numerical calculation of distribution functions

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### Introduction.

In many places in statistics (for example, Durbin and Knott (1972); Blum, Kiefer and Rosenblatt (1961); Hoeffding (1948)) one must compute the distribution function  $F$  of the infinite quadratic form  $X = \sum_{n=1}^{\infty} \lambda_n Z_n^2$ , where the coefficients satisfy  $\sum_{n=1}^{\infty} \lambda_n < \infty$  and the  $Z_n$  are independent  $N(0,1)$  random variables.

In certain applications of the theory of U-statistics (see e.g. Lee (1979)), the coefficients  $\lambda_n$  are not available in closed form but are known to be the eigenvalues of a certain integral operator. In this case the  $\lambda_n$  must be determined numerically. In other applications, the  $\lambda_n$  are explicitly known and the characteristic function  $\phi$  corresponding to  $F$ , namely  $\phi(t) = \prod_{n=1}^{\infty} (1 - 2i\lambda_n t)^{-1/2}$ , may be expressed in closed form.  $F$  may then be obtained by numerical inversion of the c.f.  $\phi$ . (For techniques of numerical inversion, see e.g. Bohman (1972), Davies (1973).) If the infinite product defining  $\phi$  cannot be written in closed form, the technique to be described may still be used to advantage.

Accordingly, it seems useful to have a method for the evaluation of  $F$  that depends on knowledge of the first  $N$  eigenvalues (arranged in descending order of magnitude) and also on the four quantities (traces)

$$S_j = \sum_{n=1}^{\infty} \lambda_n^j,$$

which may be obtained from the kernel of the integral operator without explicit determination of the  $\lambda_n$  (in the case that all the  $\lambda_n$  are not known) or by simple calculation in the case where the  $\lambda_n$  are known explicitly.

Description of the methods.

A naive approach is to approximate  $F$  by  $F^{(N)}$ , where  $F_N$  is the d.f. corresponding to the truncated c.f.

$$\phi^{(N)}(t) = \prod_{n=1}^N (1 - 2i\lambda_n t)^{-1/2}$$

$F^{(N)}$  is recovered from  $\phi^{(N)}$  by numerical inversion. However, unless the quadratic form is finite, and all eigenvalues are used, this method may lead to substantial inaccuracy. (For a discussion, see Blum, Kiefer and Rosenblatt (1961).) A much better method, advocated by Durbin and Knott (1972), is to approximate  $F$  by  $F_1$ , where  $F_1$  is the d.f. of a random variable of the form

$$X_1 = \sum_{n=1}^N \lambda_n Z_n^2 + cY$$

where  $Y$  is an  $\chi^2$  variate with  $q$  degrees of freedom, and  $c$  and  $q$  are chosen to make the first two cumulants of  $X$  and  $X_1$  agree, i.e.

$$c = S_{2,N} / S_{1,N} ,$$

$$q = S_{1,N}^2 / S_{2,N} ,$$

where  $S_{j,N} = S_j - \sum_{n=1}^N \lambda_n^j$ . One then computes the c.f. of  $X_1$ , namely

$\phi_1(t) = (1 - 2ict)^{-q/2} \prod_{n=1}^N (1 - 2i\lambda_n t)^{-1/2}$ , and recovers the approximating  $F_1$  by numerical inversion of  $\phi_1$ .

This method is effective when all  $\lambda_n$  are positive, but one expects inaccuracies if substantial numbers of the  $\lambda_n$ ,  $n > N$ , are negative. A modification of the method consists of using the approximation  $F_2$  where  $F_2$  is the d.f. of the r.v.

$$X_2 = \sum_{n=1}^N \lambda_n Z_n^2 + \alpha Y_1 + \beta Y_2$$

where  $Y_1, Y_2$  are  $\chi^2$  variates with  $q_1$  and  $q_2$  d.f., independent of each other and the  $Z_n$ . To make the first four cumulants of  $X$  and  $X_2$  coincide, the parameters are chosen to be the solutions of the equations

$$\alpha^j q_1 + \beta^j q_2 = S_{j,N} \quad j = 1, \dots, 4 .$$

Elementary algebra shows that  $\alpha$  and  $\beta$  are the roots of the quadratic

$$a^2(S_{1,N} S_{3,N} - S_{2,N}^2) + a(S_{3,N} S_{2,N} - S_{1,N} S_{4,N}) + (S_{4,N} S_{2,N} - S_{3,N}^2) = 0 .$$

Assuming  $\alpha, \beta$  to be real, we then have

$$q_1 = (S_{2,N} - \beta S_{1,N}) / \alpha(\alpha - \beta)$$

$$q_2 = (\alpha S_{1,N} - S_{2,N}) / \beta(\alpha - \beta) .$$

Assuming  $q_1, q_2$  to be positive, we approximate  $\phi$  by

$$\phi_2(t) = (1-2i\alpha t)^{-q_1/2} (1-2i\beta t)^{-q_2/2} \prod_{i=1}^N (1-2i\lambda_n t)^{-1/2} \text{ and invert } \phi_2 \text{ to obtain } F_2 .$$

The numerical inversion technique used in the program in the appendix is that of Davies (1973), which approximates the d.f.  $F_2$  at  $x$  by

$$\frac{1}{2} - \frac{\Delta}{\pi} \sum_{k=0}^K \operatorname{Im}[\phi_2((k+\frac{1}{2})\Delta) \exp -i(k+\frac{1}{2})\Delta x] / (k+\frac{1}{2})\Delta .$$

If  $\epsilon$  is the maximum permitted error in the inversion, the spacing parameter  $\Delta$  and the truncation parameter  $K$  are chosen to make

$$\max(\operatorname{Pr}[X_2 > x + 2\pi/\Delta] , \operatorname{Pr}[X_2 < x - 2\pi/\Delta]) < \epsilon/2$$

and the truncation error in the series is less than  $\epsilon/2$ . Suitable values of  $K$  are easily chosen by noting the sensitivity of the results to different values. The choice of  $\Delta$  is facilitated by the estimates

$$\max(\operatorname{Pr}[X_2 > \delta_1] , \operatorname{Pr}[X_2 < -\delta_2]) \leq \operatorname{Pr}[|X_2| > \delta]$$

where  $\delta_1, \delta_2 > 0$  and  $\delta = \min(\delta_1, \delta_2)$ . Now

$$\begin{aligned} \operatorname{Pr}[|X_2| > \delta] &\leq \operatorname{Pr}\left[\sum_{n=1}^N |\lambda_n| Z_n + |\alpha| Y_1 + |\beta| Y_2 > \delta\right] \\ &\leq \exp\left(-\frac{S_1^*}{2\lambda^*} \left(\frac{\delta}{S_1^*} - 1 - \log(\delta/S_1^*)\right)\right) \end{aligned} \quad (1)$$

where the last estimate is obtained using the methods of Blum et al. (1961)

and

$$\begin{aligned} S_1^* &= \sum_{n=1}^N |\lambda_n| + |\alpha| + |\beta| \\ \lambda^* &= \max(|\lambda_n| , n=1, \dots, N, |\alpha| , |\beta|) . \end{aligned}$$

Thus we choose  $\Delta$  so that  $x + 2\pi/\Delta$  and  $x - 2\pi/\Delta$  are respectively positive and negative, and at least  $\delta$  in magnitude, where  $\delta$  is such that (1) is less than  $\epsilon/2$  in magnitude.

Examples.

The three techniques (simple truncation, single  $\chi^2$  approximation, double  $\chi^2$  approximation) were tried on two examples with the results noted below.

Example 1. Let the coefficients  $\lambda_n$  be given by  $\lambda_{2n} = 1/\pi^2 n^2$ ,  $\lambda_{2n-1} = 1/\pi^2 n^2$ ,  $n = 1, 2, \dots$ . The characteristic function for  $t > 0$  is given by  $\phi(t) = \prod_{n=1}^{\infty} (1 - 2it/\pi^2 n^2)^{-1} = \frac{z}{\sin z}$  where  $z = (1+i)\sqrt{t}$ . The distribution function is  $F(x) = 1 + 2 \sum_{n=1}^{\infty} (-1)^n \exp(-\pi^2 n^2 x/2)$ . Table I below gives the results of using the program in the appendix.

TABLE I

Results rounded to 5 decimal places,  $\Delta = 2$ ,  $K = 200$ .  
Calculation of  $F(x)$  for selected  $x$ -values.

	$x = .2$	$x = .5$	$x = 1.0$	$x = 1.5$
Exact	.29290	.83049	.98562	.99878
Numerical Inversion	.29290	.83049	.98562	.99878
Truncation				
$N = 4$	.50949	.88694	.99041	.99919
$N = 10$	.39715	.85871	.98801	.99898
$N = 20$	.34855	.84588	.98692	.99889
Single $\chi^2$				
$N = 4$	.29228	.83054	.98562	.99878
$N = 6$	.29280	.83050	.98562	.99878
$N = 8$	.29287	.83050	.98562	.99878
Double $\chi^2$				
$N = 4$	.29290	.83049	.98562	.99878
$N = 6$	.29290	.83049	.98562	.99878
$N = 8$	.29290	.83049	.98562	.99878

Example 2. Let now the coefficients be given by  $\lambda_n = (-1)^{n-1}/\pi^2 n^2$ .

The characteristic function is for  $t > 0$

$$\phi(t) = [\cos z \overline{\sin hz / z}]^{-1/2}, \quad z = (1+i)\sqrt{t}/2.$$

Table II below gives the results of computations using the program in the appendix.

TABLE II

Results rounded to 5 decimal places,  $\Delta = 2$ ,  $K = 200$ .  
Calculation of  $F(x)$  for selected  $x$ -values.

	$x = 0.0$	$x = .5$	$x = 1.0$	$x = 1.5$
Numerical Inversion	.25506	.97564	.99844	.99989
Truncation				
$N = 4$	.27013	.97599	.99846	.99989
$N = 10$	.25820	.97570	.99844	.99989
$N = 20$	.25592	.97565	.99844	.99989
Single $\chi^2$				
$N = 4$	.25602	.97563	.99844	.99989
$N = 6$	.25557	.97563	.99844	.99989
$N = 8$	.25533	.97564	.99844	.99989
Double $\chi^2$				
$N = 4$	.25507	.97564	.99844	.99989
$N = 6$	.25506	.97564	.99844	.99989
$N = 8$	.25506	.97564	.99844	.99989

Remarks. Truncation is obviously unsatisfactory, except possibly on the extreme tails of the distributions. The single  $\chi^2$  approximation is a considerable improvement, especially when all coefficients are positive.

However, the double  $\chi^2$  approximation is clearly superior except in the tails, giving results correct to 5 decimals with only four eigenvalues in the first example and within one digit in the fifth decimal place in the second.

#### APPENDIX

The FORTRAN program below calculates either the truncation, single  $\chi^2$  or double  $\chi^2$  approximation to F. The subroutine employs the following formal parameters:

##### Input

EIG: vector of N coefficients  $\lambda_n$   
 N: dimension of EIG  
 CUM: 4-vector of quantities  $S_j$   
 X: m-vector of arguments  $x_i$  of distribution function  
 M: dimension of X  
 IND: set to 0 for truncation, 1 for single  $\chi^2$  and 2 for double  $\chi^2$  approximation  
 DELTA: spacing parameter  $\Delta$   
 KTRUNC: truncation parameter K

##### Output

F: m-vector of m function values  $F(x_i)$   
 IER: error indicator. Set to 0 if double  $\chi^2$  approximation fails.

The program should give satisfactory results in single precision on a 32-bit machine.



```

SUBROUTINE DISTRB(EIG,N,CUM,X,M,F,IND,DELTA,KTRUNC,IER)
DIMENSION EIG(N),X(M),CUM(4),SUM(4),S(4)
REAL IMAG,IMAG1
DATA PI/3.141592654/
IER=1
IF(IND.EQ.0)GO TO 7
C
C NOW SET UP FOR CHI-SQUARE APPROXIMATIONS
C
LIM=2*IND
DO 1 I=1,LIM
1 S(I)=CUM(I)
DO 3 N1=1,N
TM=1.0
DO 2 J=1,4
TM=TM*EIG(N1)
2 S(J)=S(J)-TM
3 CONTINUE
IF(IND.EQ.1)GO TO 55
C
C NOW SET UP FOR DOUBLE CHI-SQUARE APPROXIMATIONS
C
Z1=S(1)/S(2)
Z3=S(3)/S(2)
Z4=S(4)/S(2)
A=Z3*Z1-1.0
B=Z3-Z1*Z4
C=Z4-Z3*Z3
D=B*B-4.0*A*C
IF(D.GE.0.0)GO TO 5
C
C EXIT IF DOUBLE APPROXIMATION FAILS.
C
IER=0
RETURN
C
C CALCULATE PARAMETERS OF DOUBLE APPROXIMATION
C
5 AL=-(B+SQRT(D))/(2.0*A)
BE=(Z3-AL)/(1.0-AL*Z1)
XX=(1.0-BE*Z1)/(AL-BE)
Q1=S(2)*XX/AL
Q2=(Z1-XX)*S(2)/BE
IF(Q1.LE.0.0)GO TO 50
IF(Q2.GT.0.0)GO TO 7
C
C EXIT IF DOUBLE APPROXIMATION FAILS.
C
50 IER=0
RETURN
C
C CALCULATE PARAMETERS OF SINGLE APPROXIMATION
C

```

```

55  C=S(2)/S(1)
    Q=S(1)*S(1)/S(2)
C
C  BEGIN NUMERICAL INVERSION
C
7  DO 9 I=1,M
9  SUM(I)=0.0
    DO 8 K=0,KTRUNC
      T=(K+.5)*DELTA
C
C  COMPUTE C.F. OF FINITE QUADRATIC FORM
C
    REAL=1.0
    IMAG=0.0
    DO 20 L=1,N
      S3=2.0*T*EIG(L)
      T3=ATAN(-S3)
      RHO3=SQRT(SQRT(1.0+S3*S3))
      REAL1=COS(T3/2.0)/RHO3
      IMAG1=-SIN(T3/2.0)/RHO3
      REAL2=REAL*REAL1-IMAG*IMAG1
      IMAG=IMAG*REAL1+REAL*IMAG1
      REAL=REAL2
20  CONTINUE
    IF(IND.EQ.0)GO TO 10
    IF(IND.EQ.2)GO TO 11
C
C  ADJUST C.F. FOR SINGLE APPROXIMATION
C
    CALL CHISQ(T,REAL,IMAG,C,Q)
    GO TO 10
C
C  ADJUST C.F. FOR DOUBLE APPROXIMATION
C
11  CALL CHISQ(T,REAL,IMAG,AL,Q1)
    CALL CHISQ(T,REAL,IMAG,BE,Q2)
C
C  COMPLETE NUMERICAL INVERSION
C
10  DO 12 I=1,M
    SUM(I)=SUM(I)+(IMAG*COS(T*X(I))-REAL*SIN(T*X(I)))/T
12  CONTINUE
8  CONTINUE
    DO 14 I=1,M
      F(I)=.5-SUM(I)*DELTA/PI
14  CONTINUE
    RETURN
    END
    SUBROUTINE CHISQ(T,REAL,IMAG,C,Q)
C
C  SUBROUTINE MULTIPLIES THE COMPLEX NUMBER (REAL,IMAG) BY (1-C*I*T)*(-Q/2.0)
    AND RETURNS RESULT AS (REAL,IMAG).
C

```

```
REAL IMAG,IMAG1
S=-2*C*T
R=SQRT(1.0+S*S)
THETA=ATAN(S)*(-Q/2.0)
RHO=R**(-Q/2.0)
REAL1=RHO*COS(THETA)
IMAG1=RHO*SIN(THETA)
REAL2=REAL*REAL1-IMAG*IMAG1
IMAG=REAL1*IMAG+REAL*IMAG1
REAL=REAL2
RETURN
END
```

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