

## Computation of the Bivariate Normal Integral

By Z. Drezner

**Abstract.** This paper presents a simple and efficient computation for the bivariate normal integral based on direct computation of the double integral by the Gauss quadrature method.

**1. Introduction.** The probability distribution of the normalized Normal Distribution is: [1]

$$(1) \quad \Phi(h, k, \rho) = \Pr\{(x_1 < h) \cap (x_2 < k)\},$$

$$\Phi(h, k, \rho) = (2\pi\sqrt{1-\rho^2})^{-1} \int_{-\infty}^h \int_{-\infty}^k \exp\left[-\frac{x_1^2 - 2\rho x_1 x_2 + x_2^2}{2(1-\rho^2)}\right] dx_1 dx_2.$$

Substitute:

$$(2) \quad u_1 = \frac{h - x_1}{[2(1-\rho^2)]^{1/2}}; \quad u_2 = \frac{k - x_2}{[2(1-\rho^2)]^{1/2}}.$$

Define:

$$(3) \quad h_1 = \frac{h}{[2(1-\rho^2)]^{1/2}}; \quad k_1 = \frac{k}{[2(1-\rho^2)]^{1/2}}.$$

Then:

$$(4) \quad \Phi(h, k, \rho) = \frac{(1-\rho^2)^{1/2}}{\pi} \int_0^\infty \int_0^\infty \exp[-u_1^2] \exp[-u_2^2] \exp[h_1(2u_1 - h_1) + k_1(2u_2 - k_1) + 2\rho(u_1 - h_1)(u_2 - k_1)] du_1 du_2.$$

By Gauss quadrature [2]:

$$(5) \quad \Phi(h, k, \rho) \doteq \frac{(1-\rho^2)^{1/2}}{\pi} \sum_{i,j=1}^k A_i A_j f(x_i, x_j),$$

where

$$(6) \quad f(x, y) = \exp[h_1(2x - h_1) + k_1(2y - k_1) + 2\rho(x - h_1)(y - k_1)].$$

The values of  $A_i, x_i$  for  $k = 2, \dots, 15$ , can be found in [3]. If  $h, k, \rho \leq 0$ , then  $0 < f(x, y) \leq 1$ , and the error in (5) is relatively small. We will make use of the following formulae in order to calculate the double integral for  $h, k, \rho \leq 0$ .

Received February 7, 1977.

AMS (MOS) subject classifications (1970). Primary 62-04, 33A20.

Copyright © 1978, American Mathematical Society

2. **The Method.** The following formulae can be found in [1]:

$$(7) \quad \Phi(h, k, \rho) = \phi(h) + \phi(k) - 1 + \Phi(-h, -k, \rho),$$

$$(8) \quad \Phi(h, k, \rho) = \phi(k) - \Phi(-h, k, -\rho),$$

$$(9) \quad \Phi(h, k, \rho) = \phi(h) - \Phi(h, -k, -\rho),$$

where

$$(10) \quad \phi(h) = \frac{1}{2\pi} \int_{-\infty}^h \exp\left[-\frac{x^2}{2}\right] dx.$$

For  $h, k \neq 0$

$$(11) \quad \Phi(h, k, \rho) = \Phi(h, 0, \rho(h, k)) + \Phi(k, 0, \rho(k, h)) - \delta_{hk},$$

where

$$(12) \quad \rho(h, k) = \frac{(\rho h - k) \text{Sgn}(h)}{\sqrt{h^2 - 2\rho h k + k^2}}, \quad \delta_{hk} = \frac{1 + \text{Sgn}(h) \cdot \text{Sgn}(k)}{4}$$

and

$$\text{Sgn}(x) = \begin{cases} 1, & x \geq 0 \\ -1, & x < 0 \end{cases}.$$

*Algorithm.* If  $h \cdot k \cdot \rho \leq 0$ , then do one of the following:

- (a) If  $h \leq 0, k \leq 0, \rho \leq 0$  compute directly.
- (b) If  $h \leq 0, k \geq 0, \rho \geq 0$  use (9).
- (c) If  $h \geq 0, k \leq 0, \rho \geq 0$  use (8).
- (d) If  $h \geq 0, k \geq 0, \rho \leq 0$  use (7).

If  $h \cdot k \cdot \rho > 0$ , use (11). Note that every computation of  $\Phi$  will now satisfy  $h \cdot k \cdot \rho = 0$ , (since the new  $k$  equals 0).

3. **Results.** An advantage of this method is that for every  $\rho$  (even close to 1) there is no convergence problem. In Table 1 we present results of the average run time and maximum-error for various values of  $k$  in (5).

Low values of the exponent in (6) cause  $f(x, y)$  to vanish. To save computational effort, if the exponent is lower than the values in Column 4, Table 1, we assume that  $f(x, y)$  is zero. The values in the table have been set such that maximum-error remains the same to two significant digits. In Column 5 we present the reduced run time. In order to compare with existing results [4], [5] we take  $k = 5$ . Note here that in  $k = 3, 4, 5$  a further reduction of 0.7 m.s. can be achieved by using the approximation in [6] for the error function instead of using the function erf.

By a regression on Column 3 we can deduce that computations outside the double integration are the same for every  $k$  and require approximately 1.6 m.s., so for  $k = 5$  and the approximated error function the average computation time is 2.2 m.s. The double integration requires an average 1.3 m.s.

TABLE 1. *Results*

$k$	maximum error	run time ( $10^{-3}$ sec)	limit of exponent	reduced run time
3	$1.1 \times 10^{-4}$	3.0	- 8	1.8
4	$8.1 \times 10^{-6}$	4.1	-10	2.2
5	$5.5 \times 10^{-7}$	5.5	-12	2.9
6	$3.8 \times 10^{-8}$	7.2	-15	3.6
7	$3.0 \times 10^{-9}$	9.2	-17	4.6
8	$2.2 \times 10^{-10}$	11.5	-20	6.0
9	$1.5 \times 10^{-11}$	14.2	-22	7.5
10	$1.1 \times 10^{-12}$	17.0	-25	9.4

All time data are based on a CDC/6400.

Faculty of Business  
McMaster University  
Hamilton, Ontario, Canada

1. NORMAN L. JOHNSON & SAMUEL KOTZ, *Distribution in Statistics: Continuous Multivariate Distributions*, Wiley, New York, 1972, pp. 93-96.
2. A. H. STROUD & D. SECREST, *Gaussian Quadrature Formulas*, Prentice-Hall, Englewood Cliffs, N. J., 1966.
3. N. M. STEEN, G. O. BYRNE & E. M. GELBARD, "Gaussian Quadratures," *Math. Comp.*, v. 23, 1969, pp. 661-671.
4. R. R. SOWDEN & J. R. ASHFORD, "Computation of the bivariate normal integral," *Appl. Statist.*, v. 18, 1969, pp. 169-180.
5. D. E. AMOS, "On computation of the bivariate normal distribution," *Math. Comp.*, v. 23, 1969, pp. 655-659.
6. Y. L. LUKE, *Mathematical Functions and Their Approximations*, Academic Press, New York, 1975, pp. 123-124.