

$$X_n = \left( \sum_{i=1}^n U_i - \frac{n}{2} \right) \left( \frac{n}{12} \right)^{-1/2}$$

will be distributed asymptotically as a normal random deviate. When  $n=12$ , the maximum errors made in the normal deviate are  $9 \times 10^{-3}$  for  $|X| < 2$ ,  $9 \times 10^{-1}$  for  $2 < |X| < 3$ . An improvement can be made by taking a polynomial function of  $X_n$  (say)

$$X_n^* = X_n \sum_{s=0}^k a_{2s} X_n^{2s}$$

as the normal deviate where  $a_{2s}$  are suitable coefficients. These coefficients may be calculated using (say) Chebyshev polynomials or simply by making the asymptotic random deviate agree with the correct normal deviate at certain specified points. When  $n=12$ , the maximum error in the normal deviate is  $8 \times 10^{-4}$  using the coefficients

$$\begin{aligned} * a_0 &= 9.8746 & * a_6 &= (-7) - 5.102 \\ * a_2 &= (-3)3.9439 & * a_8 &= (-7)1.141 \\ * a_4 &= (-5)7.474 \end{aligned}$$

(3) *Direct method*: Generate a pair of uniform deviates  $(U_1, U_2)$ . Then

$$X_1 = (-2 \ln U_1)^{1/2} \cos 2\pi U_2,$$

$X_2 = (-2 \ln U_1)^{1/2} \sin 2\pi U_2$  will be a pair of independent normal random deviates with mean zero and unit variance. This procedure can be modified by calculating  $\cos 2\pi U$  and  $\sin 2\pi U$  using an acceptance rejection method; e.g., (1) generate  $(U_1, U_2)$ ; (2) if  $(2U_1 - 1)^2 + (2U_2 - 1)^2 \leq 1$  generate a third uniform deviate  $U_3$ , otherwise reject the pair and start over; (3) calculate  $y_1 = (-\ln u_3)^{1/2} \frac{u_1^2 - u_2^2}{u_1^2 + u_2^2}$ ,  $y_2 = \pm 2(-\ln u_3)^{1/2} \frac{u_1 u_2}{u_1^2 + u_2^2}$  ( $\pm$  random). Both  $y_1$  and  $y_2$  are the desired random deviates.

(4) *Acceptance-rejection method*: 1) Generate a pair of uniform deviates  $(U_1, U_2)$ ; 2) compute  $x = -\ln u_1$ ; 3) if  $e^{-\frac{1}{2}(x-1)^2} \geq u_2$  (or equivalently  $(x-1)^2 \leq -2(\ln u_2)$ ) accept  $x$ , otherwise reject the

pair and start over. The quantity will be the required normal deviate with mean zero and unit variance.

**b. Bivariate normal distribution**

Let  $\{X_1, X_2\}$  be a pair of independent normal deviates with mean zero and unit variance. Then  $\{X_1, \rho X_1 + (1 - \rho^2)^{1/2} X_2\}$  represent a pair of deviates from a bivariate normal distribution with zero means, unit variances, and correlation coefficient  $\rho$ .

**c. Exponential distribution**

(1) *Inverse method*: Since  $F(x) = e^{-x/\theta}$ ,  $X = -\theta \ln U$  will be a deviate from the exponential distribution with parameter  $\theta$ .

(2) *Acceptance-rejection method*: 1) Generate a pair of independent uniform deviates  $(U_0, U_1)$ ; 2) if  $U_1 < U_0$  generate a third value  $U_2$ ; 3) if  $U_1 + U_2 < U_0$  generate a fourth value  $U_3$ , etc.; 4) continue generating uniform deviates until an  $n$  is obtained such that  $U_1 + U_2 + \dots + U_{n-1} < U_0 < U_1 + \dots + U_n$ ; 5) if  $n$  is even reject the procedure and start a fresh trial with a new value of  $U_0$ , otherwise if  $n$  is odd take  $X = \theta U_0$  as the desired deviate; 6) in general if  $t$  is the number of trials until an acceptable sequence is obtained  $X = \theta(t + U_0)$ . The random deviates produced in this way follow an exponential distribution with parameter  $\theta$ . One can expect to generate approximately six uniform deviates for every exponential deviate.

(3) *Discrete Distribution Method*: Let  $Y$  and  $n$  be discrete random variables with point probabilities

$$\begin{aligned} * Pr\{Y=r\} &= (e-1)e^{-(r+1)} \quad r=0, 1, 2, \dots \\ Pr\{n=s\} &= [s!(e-1)]^{-1} \quad s=1, 2, 3, \dots \end{aligned}$$

Then  $X = Y + \min(U_1, U_2, \dots, U_n)$  will follow an exponential distribution. The average value of  $n$  is 1.58 so that one needs, on the average, only 1.58  $u$ 's from which the minimum is selected.

**26.9. Use and Extension of the Tables**

**Use of Probability Function Inequalities**

**Example 1.** Let  $X$  be a random variable with finite mean and variance equal to  $m$  and  $\sigma^2$ , respectively. Use the inequalities for probability functions 26.1.37, 40, 41 to place lower bounds on

$$A(t) = F(t) - F(-t) = P \left\{ \frac{|X - m|}{\sigma} \leq t \right\}$$

for  $t=1(1)4$ .

Lower bounds on $A(t) = F(t) - F(-t)$				Remarks
$t=1$	$2$	$3$	$4$	
0	.7500	.8889	.9375	no knowledge of $F(t)$ ; 26.1.37
.5556	.8889	.9506	.9722	$F(t)$ is unimodal and continuous; 26.1.40
0	.8182	.9697	.9912	$F(t)$ is such that $\mu_4=3$ ; 26.1.41

\*See page II.