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Handbook of Mathematical Functions

With

Formulas, Graphs, and Mathematical Tables

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26. Probability Functions

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26. Probability Functions

Mathematical Properties³

26.1. Probability Functions: Definitions and Properties

Univariate Cumulative Distribution Functions

A real-valued function $F(x)$ is termed a (univariate) cumulative distribution function (c.d.f.) or simply distribution function if

- i) $F(x)$ is non-decreasing, i.e., $F(x_1) \leq F(x_2)$ for $x_1 \leq x_2$
- ii) $F(x)$ is everywhere continuous from the right, i.e., $F(x) = \lim_{\epsilon \rightarrow 0^+} F(x + \epsilon)$
- iii) $F(-\infty) = 0, F(\infty) = 1$.

The function $F(x)$ signifies the probability of the event " $X \leq x$ " where X is a random variable, i.e., $Pr\{X \leq x\} = F(x)$, and thus describes the c.d.f. of X . The two principal types of distribution functions are termed *discrete* and *continuous*.

Discrete Distributions: Discrete distributions are characterized by the random variable X taking on an enumerable number of values . . . , x_{-1}, x_0, x_1, \dots with point probabilities

$$p_n = Pr\{X = x_n\} \geq 0$$

which need only be subject to the restriction

$$\sum_n p_n = 1.$$

The corresponding distribution function can then be written

$$26.1.1 \quad F(x) = Pr\{X \leq x\} = \sum_{x_n \leq x} p_n$$

³ *Comment on notation and conventions.*

a. We follow the customary convention of denoting a random variable by a capital letter, i.e., X , and using the corresponding lower case letter, i.e., x , for a particular value that the random variable assumes.

b. For statistical applications it is often convenient to have tabulated the "upper tail area," $1 - F(x)$, or the c.d.f. for $|X|$, $F(x) - F(-x)$, instead of simply the c.d.f. $F(x)$. We use the notation P to indicate the c.d.f. of X , $Q = 1 - P$ to indicate the "upper tail area" and $A = P - Q$ to denote the c.d.f. of $|X|$. In particular we use $P(x)$, $Q(x)$, and $A(x)$ to denote the corresponding functions for the normal or Gaussian probability function, see 26.2.2-26.2.4. When these distributions depend on other parameters, say θ_1 and θ_2 , we indicate this by writing $P(x|\theta_1, \theta_2)$, $Q(x|\theta_1, \theta_2)$, or $A(x|\theta_1, \theta_2)$. For example the chi-square distribution 26.4 depends on the parameter ν and the tabulated function is written $Q(x^2|\nu)$.

where the summation is over all values of x for which $x_n \leq x$. The set $\{x_n\}$ of values for which $p_n > 0$ is termed the domain of the random variable X . A discrete distribution of a random variable is called a *lattice distribution* if there exist numbers a and $b \neq 0$ such that every possible value of X can be represented in the form $a + bn$ where n takes on only integral values. A summary of some properties of certain discrete distributions is presented in 26.1.19-26.1.24.

Continuous Distributions. Continuous distributions are characterized by $F(x)$ being absolutely continuous. Hence $F(x)$ possesses a derivative $F'(x) = f(x)$ and the c.d.f. can be written

$$26.1.2 \quad F(x) = Pr\{X \leq x\} = \int_{-\infty}^x f(t) dt.$$

The derivative $f(x)$ is termed the *probability density function* (p.d.f.) or *frequency function*, and the values of x for which $f(x) > 0$ make up the domain of the random variable X . A summary of some properties of certain selected continuous distributions is presented in 26.1.25-26.1.34.

Multivariate Probability Functions

The real-valued function $F(x_1, x_2, \dots, x_n)$ defines an n -variate cumulative distribution function if

- i) $F(x_1, x_2, \dots, x_n)$ is a non-decreasing function for each x_i
- ii) $F(x_1, x_2, \dots, x_n)$ is continuous from the right in each x_i ; i.e., $F(x_1, x_2, \dots, x_n) = \lim_{\epsilon \rightarrow 0^+} F(x_1, \dots, x_i + \epsilon, \dots, x_n)$
- iii) $F(x_1, x_2, \dots, x_n) = 0$ when any $x_i = -\infty$; $F(\infty, \infty, \dots, \infty) = 1$. The function $F(x_1, x_2, \dots, x_n)$ signifies the probability of the event $X_1 \leq x_1, X_2 \leq x_2, \dots, X_n \leq x_n$ where X_1, X_2, \dots, X_n is a set of n random variables.

Thus $Pr\{X_1 \leq x_1, X_2 \leq x_2, \dots, X_n \leq x_n\} = F(x_1, x_2, \dots, x_n)$. The two principal types of n -variate distribution functions termed *discrete* and *continuous*, are defined in a manner similar to the corresponding cases for the univariate distribution function.

Series Expansion

26.6.22

$$P(F'|v_1, v_2, \lambda) = e^{-\frac{\lambda}{2}(1-x)} x^{\frac{1}{2}(v_1+v_2-2)} \sum_{i=0}^{\frac{v_2-1}{2}} T_i \quad (v_2 \text{ even})$$

where

$$T_0 = 1$$

$$T_1 = \frac{1}{2} (v_1 + v_2 - 2 + \lambda x) \frac{1-x}{x}$$

$$T_i = \frac{1-x}{2i} [(v_1 + v_2 - 2i + \lambda x) T_{i-1} + \lambda(1-x) T_{i-2}]$$

$$x = \frac{v_2}{v_1 F' + v_2}$$

Limiting Forms

26.6.23

$$\lim_{v_2 \rightarrow \infty} P(F'|v_1, v_2, \lambda) = P(\chi'^2|v, \lambda), \quad \chi'^2 = v_1 F', \quad v = v_1$$

26.6.24

$$\lim_{v_1 \rightarrow \infty} P(F'|v_1, v_2, \lambda) = Q(\chi^2|v), \quad \chi^2 = \frac{v_2(1+c^2)}{F'}$$

where $\lambda/v_1 \rightarrow c^2$ as $v_1 \rightarrow \infty$.

Approximations to the Non-Central F-Distribution

26.6.25 $P(F'|v_1, v_2, \lambda) \approx P(x_1)$, $(v_1 \text{ and } v_2 \text{ large})$

where

$$x_1 = \frac{F' - \frac{v_2(v_1 + \lambda)}{v_1(v_2 - 2)}}{\frac{v_2}{v_1} \left[\frac{2}{(v_2 - 2)(v_2 - 4)} \left\{ \frac{(v_1 + \lambda)^2}{v_2 - 2} + v_1 + 2\lambda \right\} \right]^{\frac{1}{2}}}$$

26.6.26

$$P(F'|v_1, v_2, \lambda) \approx P(F|v_1^*, v_2),$$

$$F = \frac{v_1}{v_1 + \lambda} F', \quad v_1^* = \frac{(v_1 + \lambda)^2}{v_1 + 2\lambda}$$

26.6.27

$$P(F'|v_1, v_2, \lambda) \approx P(x_2),$$

$$x_2 = \frac{\left[\frac{v_1 F'}{(v_1 + \lambda)} \right]^{1/3} \left[1 - \frac{2}{9v_2} \right] - \left[1 - \frac{2(v_1 + 2\lambda)}{9(v_1 + \lambda)^2} \right]}{\left[\frac{2}{9} \frac{v_1 + 2\lambda}{(v_1 + \lambda)^2} + \frac{2}{9v_2} \left(\frac{v_1}{v_1 + \lambda} F' \right)^{2/3} \right]^{\frac{1}{2}}}$$

26.7. Student's *t*-Distribution

If X is a random variable following a normal distribution with mean zero and variance unity, and χ^2 is a random variable following an independent chi-square distribution with ν degrees of freedom, then the distribution of the ratio $\frac{X}{\sqrt{\chi^2/\nu}}$

is called Student's *t*-distribution with ν degrees of freedom. The probability that $\frac{X}{\sqrt{\chi^2/\nu}}$ will be less in absolute value than a fixed constant t is

26.7.1

$$A(t|\nu) = P\left\{ \left| \frac{X}{\sqrt{\chi^2/\nu}} \right| \leq t \right\}$$

$$= \left[\sqrt{\nu} B\left(\frac{1}{2}, \frac{\nu}{2}\right) \right]^{-1} \int_{-t}^t \left(1 + \frac{x^2}{\nu}\right)^{-\frac{\nu+1}{2}} dx$$

$$= 1 - 2I_x\left(\frac{\nu}{2}, \frac{1}{2}\right), \quad (0 \leq t < \infty)$$

where

$$x = \frac{t}{\sqrt{1+t^2}}$$

Statistical Properties

26.7.2

mean: $m = 0$

variance: $\sigma^2 = \frac{\nu}{\nu-2}$ ($\nu > 2$)

skewness: $\gamma_1 = 0$

excess: $\gamma_2 = \frac{6}{\nu-4}$ ($\nu > 4$)

moments:

$$\mu_{2n} = \frac{1 \cdot 3 \dots (2n-1) \nu^n}{(\nu-2)(\nu-4) \dots (\nu-2n)} \quad (\nu > 2n)$$

$$\mu_{2n+1} = 0$$

characteristic function:

$$\phi(t) = E \left[\exp \left(it \frac{X}{\sqrt{\chi^2/\nu}} \right) \right] = \frac{\left(\frac{|t|}{2\sqrt{\nu}} \right)^{\nu/2}}{\pi \Gamma(\nu/2)} Y_{\frac{\nu}{2}} \left(\frac{|t|}{\sqrt{\nu}} \right)$$

Series Expansions

$$\left(\theta = \arctan \frac{t}{\sqrt{\nu}} \right)$$

26.7.3

$$A(t|\nu) = \begin{cases} \frac{2}{\pi} \left\{ \theta + \sin \theta \left[\cos \theta + \frac{2}{3} \cos^3 \theta + \dots \right. \right. \\ \left. \left. + \frac{2 \cdot 4 \dots (\nu-3)}{1 \cdot 3 (\nu-2)} \cos^{\nu-2} \theta \right] \right\} & (\nu > 1 \text{ and odd}) \\ \frac{2}{\pi} \theta & (\nu = 1) \end{cases}$$

26.7.4

$$A(t|\nu) = \sin \theta \left\{ 1 + \frac{1}{2} \cos^2 \theta + \frac{1 \cdot 3}{2 \cdot 4} \cos^4 \theta + \dots \right. \\ \left. + \frac{1 \cdot 3 \cdot 5 \dots (\nu-3)}{2 \cdot 4 \cdot 6 (\nu-2)} \cos^{\nu-2} \theta \right\} \quad (\nu \text{ even})$$

Asymptotic Expansion for the Inverse Function

If $A(t|\nu)=1-2p$ and $Q(x_p)=p$, then

26.7.5

$$t_p \sim x_p + \frac{g_1(x_p)}{\nu} + \frac{g_2(x_p)}{\nu^2} + \frac{g_3(x_p)}{\nu^3} + \dots$$

$$g_1(x) = \frac{1}{4}(x^3 + x)$$

$$g_2(x) = \frac{1}{96}(5x^5 + 16x^3 + 3x)$$

$$g_3(x) = \frac{1}{384}(3x^7 + 19x^5 + 17x^3 - 15x)$$

$$g_4(x) = \frac{1}{92160}(79x^9 + 776x^7 + 1482x^5 - 1920x^3 - 945x)$$

Limiting Distribution

26.7.6

$$\lim_{\nu \rightarrow \infty} A(t|\nu) = \frac{1}{\sqrt{2\pi}} \int_{-t}^t e^{-x^2/2} dx = A(t)$$

Approximation for Large Values of t and $\nu \leq 5$

26.7.7 $A(t|\nu) \approx 1 - 2 \left\{ \frac{a_\nu}{t^\nu} + \frac{b_\nu}{t^{\nu+1}} \right\}$

ν	1	2	3	4	5
a_ν	.3183	.4991	1.1094	3.0941	9.948
b_ν	.0000	.0518	-.0460	-2.756	-14.05

Numerical Methods

26.8. Methods of Generating Random Numbers and Their

Random digits are digits generated by repeated independent drawings from the population 0, 1, 2, . . . , 9 where the probability of selecting any digit is one-tenth. This is equivalent to putting 10 balls, numbered from 0 to 9, into an urn and drawing one ball at a time, replacing the ball after each drawing. The recorded set of numbers forms a collection of random digits. Any group of n successive random digits is known as a *random number*.

Several lengthy tables of random digits are available (see references). However, the use of random numbers in electronic computers has resulted in a need for random numbers to be generated in a completely deterministic way. The numbers so generated are termed pseudo-random numbers. The quality of pseudo-random numbers is determined by subjecting the numbers to several statistical tests, see [26.55], [26.56]. The purpose of these statistical tests is to detect any properties of the pseudo-random numbers which are different from the (conceptual) properties of random numbers.

^o The authors wish to express their appreciation to Professor J. W. Tukey who made many penetrating and helpful suggestions in this section.

Approximation for Large ν

26.7.8 $A(t|\nu) \approx 2P(x) - 1$, $x = \frac{t \left(1 - \frac{1}{4\nu}\right)}{\sqrt{1 + \frac{t^2}{2\nu}}}$

Non-Central t-Distribution

26.7.9

$$P(t'|\nu, \delta) = \frac{1}{\sqrt{\nu} B\left(\frac{1}{2}, \frac{\nu}{2}\right)} \int_{-\infty}^{t'} \left(\frac{\nu}{\nu+x^2}\right)^{\frac{\nu+1}{2}} e^{-\frac{1}{2} \frac{\nu \delta^2}{\nu+x^2}} H_{\nu} \left(\frac{-\delta x}{\sqrt{\nu+x^2}}\right) dx$$

$$= 1 - \sum_{j=0}^{\infty} e^{-\delta^2/2} \frac{(\delta^2/2)^j}{j!} I_x \left(\frac{1}{2} + j, \frac{\nu}{2}\right), \quad x = \frac{\nu}{\nu+t'^2}$$

where δ is termed the non-centrality parameter.

Approximation to the Non-Central t-Distribution

26.7.10

$$P(t'|\nu, \delta) \approx P(x) \quad \text{where } x = \frac{t' \left(1 - \frac{1}{4\nu}\right) - \delta}{\left(1 + \frac{t'^2}{2\nu}\right)^{\frac{1}{2}}}$$

$$P(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-t^2/2} dt$$

Normal (Gaussian) probability function

Experience has shown that the most practical method is the most practical method of generating pseudo-random numbers on a computer. The method of pseudo-random numbers is based on a sequence of numbers $n=0, 1, 2, \dots$. The method of generating pseudo-random numbers is based on a sequence of numbers $n=0, 1, 2, \dots$. The method of generating pseudo-random numbers is based on a sequence of numbers $n=0, 1, 2, \dots$.

$$X_{n+1} = aX_n + b \pmod{T}$$

where b and T are relatively prime. The choice of T is determined by the capacity and base of the computer; a and b are chosen so that: (1) the resulting sequence $\{X_n\}$ possesses the desired statistical properties of random numbers, (2) the period of the sequence is as long as possible, and (3) the speed of generation is fast. A guide for choosing a and b is to make the correlation between the numbers be near zero, e.g., the correlation between X_n and X_{n+1} is

$$\rho_1 = \frac{1 - 6 \frac{b_1}{T} \left(1 - \frac{b_1}{T}\right)}{a_1} + e$$

where

$$a_1 = a^2 \pmod{T}$$

$$b_1 = (1 + a + a^2 + \dots + a^{n-1})b \pmod{T}$$

$$|e| < a_1/T$$