

# SAMPLE SIZE REQUIRED FOR ESTIMATING THE VARIANCE WITHIN $d$ UNITS OF THE TRUE VALUE<sup>1</sup>

BY FRANKLIN A. GRAYBILL AND TERRENCE L. CONNELL

*Colorado State University*

**1. Introduction.** The problem of estimating the variance ( $\sigma^2$ ) of a normal density arises in many experimental situations. J. A. Greenwood and M. M. Sandomire [3] have presented a means of obtaining the sample size required to estimate the variance of a normal density within a given per cent of its true value. An investigator may prefer instead to estimate the variance within a given number of units. This paper will provide a two step sampling procedure to solve that problem.

Assume a preliminary sample of size  $m$ ;  $z_1, z_2, \dots, z_m$ , is taken from a normal density with variance  $\sigma^2$ . The unbiased estimator of the variance  $s_m^2$  is computed by the formula  $s_m^2 = (m - 1)^{-1} \sum (z_i - \bar{z})^2$ , and  $d$  and  $1 - \alpha$  are specified in advance. It is desired to determine  $n$ , on the basis of the preliminary sample, such that

$$(1.1) \quad P[|s_{n+1}^2 - \sigma^2| < d] > 1 - \alpha$$

where  $s_{n+1}^2$  is equal to  $(1/n) \sum_{i=1}^{n+1} (y_i - \bar{y})^2$  and where  $y_1, y_2, \dots, y_{n+1}$  is a random sample of size  $n + 1$ , from a normal density with variance  $\sigma^2$ .

Table I in Section 3 provides the sample size  $n + 1$ , such that (1.1) is true, for  $1 - \alpha = .90, .95, .99$ ;  $m = 5, 10, 15, 20, 50, 100, 200, 500, 1000$ . The only other known method for solving this problem is given in [1], which requires the use of Tchebycheff's inequality. It can be shown that the method presented in this paper provides a significantly smaller second sample size than does [1]. For some comparisons with [1], see Table III.

**2. Solution.** Equation (1.1) may be written as

$$\begin{aligned} P[|s_{n+1}^2 - \sigma^2| < d] &= E_n \{ P[(1 - a) < v < (1 + a) | n] \} \\ &= \int_1^\infty g(n) \int_{1-a}^{1+a} f_1(v | n) dv dn \end{aligned}$$

where  $E_n$  is expectation with respect to  $n$ ;  $a = d/\sigma^2$ ;  $v = s_{n+1}^2/\sigma^2$ ;  $g(\cdot)$  is the density of  $n$ , and  $f_1(\cdot | n)$  is the density of a chi-square variable divided by  $n$ , its degrees of freedom. We shall restrict  $n$  such that  $n \geq 1$ . By definition

$$\begin{aligned} f_1(v | n) &= [(n/2)^{(n/2)} / \Gamma(n/2)] v^{(n/2)-1} e^{-(n/2)v}, & 0 < v < \infty \\ &= 0, & -\infty < v \leq 0. \end{aligned}$$

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