

# A NOTE ON ESTIMATION OF $\mu^2$ IN NORMAL DENSITY

By

V.N. RAO AND J. SINGH

*School of Studies in Statistics,  
Vikram University, Ujjain-456 010*

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## SUMMARY

Let  $x_1, x_2, \dots, x_n$  be a random sample of size  $n$  from a normal population having mean  $\mu$  and variance  $C^2\mu^2$ . An estimator  $T_1^*$  of  $\mu^2$  has been developed which is biased and has smaller mean-squared error (MSE) than the usual unbiased estimator  $T = \bar{x}^2 - \frac{s^2}{n}$ ,

## INTRODUCTION

Let  $x_1, x_2, \dots, x_n$  be a random sample of size  $n$  from a normal population having mean  $\mu$  and variance  $C^2\mu^2$ . In normal population the usual unbiased estimator for estimating  $\mu^2$  is

$$T = \bar{x}^2 - \frac{s^2}{n}.$$

Searls [1] proposed an estimator of the population mean, where he utilized the known coefficient of variation. Such an estimator is although biased, has smaller mean-squared error than the usual unbiased estimator  $\bar{x}$ . Singh and Pandey [2] developed estimators of the population variance  $\sigma^2$  of a normal population utilizing known coefficient of variation. Singh, Pandey and Hirano [3] considered a general population and suggested an estimator of the population variance  $\sigma^2$  utilizing the known coefficient of kurtosis. In the present note we have tried to improve the estimator  $T$  by utilising a known value of coefficient of variation. We have considered an estimator

$$T_1 = \alpha_1 \bar{x}^2 + \alpha_2 \frac{s^2}{n},$$

where  $\alpha_1$  and  $\alpha_2$  are unknown scalars and are so determined that the mean-squared error of  $T_1$  is minimum. The estimator thus obtained is denoted by  $T_1^*$ . Obviously  $T_1^*$  is a better estimator than the usual unbiased estimator  $T$ .

ESTIMATOR  $T_1^*$  AND ITS PROPERTIES

The proposed estimator is  $T_1 = \alpha_1 \bar{x}^2 + \alpha_2 \frac{s^2}{n}$  which involves two unknown scalars  $\alpha_1$  and  $\alpha_2$ . We determine the scalars to minimize the MSE of  $T_1$ . We have

$$MSE(T_1) = \alpha_1^2 v(\bar{x}^2) + \frac{\alpha_2^2}{n^2} v(s^2) + \left\{ \alpha_1 \left( \mu^2 + \frac{\sigma^2}{n} \right) + \alpha_2 \frac{\sigma^2}{n} - \mu^2 \right\}^2 \quad \dots (1)$$

The values of  $\alpha_1$  and  $\alpha_2$  which minimize mean-squared error of  $T_1$  are

$$\alpha_1 = \frac{n(n+C^2)}{C^4 \left[ \frac{\beta_2+2n-3}{n} + \frac{4\sqrt{\beta_1}}{C} + \frac{4n}{C^2} \right] + C^2(n-1) \left[ \frac{C^2(\beta_2+2n-3)+4nc\sqrt{\beta_1+4n^2}}{\beta_2(n-1)+(3-n)} \right] + (n+C^2)^2} \quad \dots (2)$$

and

$$\alpha_2 = \frac{n}{\left\{ \left[ C^2 \frac{\beta_2}{n} + \frac{3-n}{n(n-1)} \right] + \frac{(n+C^2)^2}{C^2(n-1)} \left[ \frac{\beta_2(n-1)+(3-n)}{C^2(\beta_2+2n-3)+4nc\sqrt{\beta_1+4n^2}} \right] + 1 \right\}} \quad \dots (3)$$

where  $C^2 = \sigma^2/\mu^2$ .

For normal populations  $\beta_1=0$  and  $\beta_2=3$ , so from (2) and (3), we get

$$\alpha_1 = \frac{n(n+C^2)}{C^2(n+1)(C^2+2n) + (n+C^2)^2} \quad \dots (4)$$

and

$$\alpha_2 = \frac{n(n-1)(C^2+2n)}{C^2(n+1)(C^2+2n) + (n+C^2)^2} \quad \dots (5)$$

Therefore, if we assume the coefficient of variation to be known, the proposed estimator is

$$T_1^* = \frac{1}{C^2(n+1)(C^2+2n) + (n+C^2)^2} [n(n+C^2)\bar{x}^2 + (n-1)(C^2+2n)s^2] \quad \dots (6)$$

From expression (1), we obtain

$$MSE(T_1^*) = \frac{2C^2\mu^4(C^2+2n)}{C^2(n+1)(C^2+2n) + (n+C^2)^2} \quad \dots (7)$$

Since  $T$  is a special case of  $T_1$  with  $\alpha_1=1$  and  $\alpha_2=-1$ , the proposed estimator  $T_1^*$  is better than the usual unbiased estimator  $T$ . If  $\alpha_1 = -\alpha_2 = M$ , the proposed estimator will become

$$T_2 = M \left( \bar{x}^2 - \frac{s^2}{n} \right)$$

The value of  $M$  which minimizes the mean-squared error of  $T_2$  is

$$M = \frac{1}{\frac{2C^4}{n(n-1)} + \frac{4C^2}{n} + 1}$$

Since  $T_2$  is a sub-class of  $T_1$ , the estimator  $T_1^*$  is better than the estimator  $T_2$ , and since the unbiased estimator  $T$  is a special case of  $T_2$  with  $M=1$ , the proposed estimator  $T_2$  is better estimator than the usual unbiased estimator  $T$ .

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