

# Effects of Changing Weighing Factor in a Two Stage Shrinkage Testimator for the Mean of an Exponential Distribution<sup>1)</sup>

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## Abstract

Two stage shrinkage testimator is a kind of adaptive estimators based on a test on an initial estimate of parameter. Since weighing factor plays an important roll in assessing the properties of testimator, its choice is extremely crucial in two stage testimation. Adke, Waikar and Schuurmann(1987) proposed a testimator for the mean of an exponential distribution defined with their own weighing factor. Two alternative testimators obtained using changed weighing factors are presented, and their Mean squared error(MSE) formulae are provided in this paper. Their properties are compared with those of existing one by means of MSE.

## 1. Introduction

Two stage shrinkage testimation is an estimation procedure that incorporates the results of a preliminary test on an available initial estimate of parameter (Hogg(1974), Katti(1962), and Waikar and Katti(1971)). Waikar, Schuurmann and Raghunathan(1984) developed a testimation procedure for the mean of a normal distribution. Later, Adke, Waikar and Schuurmann(1987) extended their results to the testimation of the mean of an exponential distribution. In both cases mentioned above, a weighing factor is used in defining testimator and it plays an important role in determining the properties of testimator. Hence, the choice of weighing factor is an important problem in two stage testimation. This paper concentrates on that problem in the testimation of the mean of an exponential distribution. Two changed weighing factors used in defining testimators are introduced and their MSE's are compared with that of Adke, Waikar and Schuurmann.

This paper is composed of five sections including present one. Two stage shrinkage testimator for the mean of an exponential distribution, proposed by Adke, Waikar and Schuurmann, is briefly reviewed in section 2. In section 3, two changed weighing factors are introduced, and two testimators obtained using changed weighing factors are defined. Section 4 is devoted to the comparison of two alternative testimators with Adke, Waikar and

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Schuurmann's. MSE formulae of two alternative estimators are presented in section 4, and tables containing the relative efficiencies (to usual single sample mean) of three estimators (Adke, Waikar and Schuurmann's, & two alternative ones) are also contained in that section for comparison purpose. Final section is devoted to some concluding remarks.

## 2. Two stage estimator for the mean of an exponential distribution

This section is partly reproduced from Moon(1998, Section 2) to introduce two stage estimation procedure and some notation which is necessary in this paper. Let  $X$  be a random variable following an exponential distribution with mean  $\theta$ . Adke, Waikar and Schuurmann proposed a two stage shrinkage estimator of  $\theta$  based on an initial estimate  $\theta_0$  of  $\theta$ , which is defined as follows:

Step 1: Obtain  $n_1$  first stage samples( $X_{1i}, i = 1, 2, \dots, n_1$ ), and test

$$H_0: \theta = \theta_0 \text{ vs. } H_1: \theta \neq \theta_0$$

at significance level  $\alpha$  using the first stage sample mean  $\bar{X}_1$ .

Step 2: If  $H_0$  is accepted, estimator is defined as

$$\hat{\theta} = w \bar{X}_1 + (1-w) \theta_0 \quad 0 < w < 1.$$

If  $H_0$  is rejected, obtain  $n_2$  second stage samples( $X_{2i}, i = 1, 2, \dots, n_2$ )

and define estimator as the combined sample mean, that is

$$\hat{\theta} = (n_1 \bar{X}_1 + n_2 \bar{X}_2) / (n_1 + n_2).$$

UMPU test for testing above hypothesis is given by:

$$\text{Reject } H_0 \text{ if } Z_1 = \sum_{i=1}^{n_1} X_{1i} < k_1 \text{ or if } Z_1 > k_2,$$

where  $k_1$  and  $k_2$  are chosen to satisfy

$$\begin{aligned} 1 - \alpha &= \gamma(n_1, k_2/\theta_0) - \gamma(n_1, k_1/\theta_0) \\ &= \gamma(n_1 + 1, k_2/\theta_0) - \gamma(n_1 + 1, k_1/\theta_0) \end{aligned} \quad (2.1)$$

and  $\gamma(a, x)$  represents a usual incomplete gamma function. They defined the weighing factor  $w$  as  $w = |Z_1 - n_1 \theta_0| / (k_2 - k_1)$ . They also derived MSE of estimator and defined its relative efficiency to usual single sample mean(See Adke, Waikar and Schuurmann, p.1824-1828, for detailed formula).

### 3. Testimators with changed weighing factors

As was mentioned in previous sections, the weighing factor plays an important role in testimator. Adke, Waikar and Schuurmann defined weighing factor of their testimator as  $w = |Z_1 - n_1\theta_0| / (k_2 - k_1)$  so that a higher weight is given to  $\theta_0$  when  $\bar{X}_1$  is closer to  $\theta_0$ . For notational convenience, let's denote their testimator and weighing factor as  $\hat{\theta}_1$  and  $w_1$ , respectively. In this section, two alternative testimators (denoted as  $\hat{\theta}_2$  and  $\hat{\theta}_3$ ) defined with changed weighing factors (denoted as  $w_2$  and  $w_3$ ) are introduced.

An weighing factor  $w_1$  was chosen to give higher weight to  $\bar{X}_1$  when the information contained in first stage sample does not support an initial estimate, and vice versa. However, higher weight is not high enough when test statistic  $Z_1$  is within acceptance region but takes value close to critical values(say,  $k_1$  and  $k_2$ ). That is,  $w_1$  takes value much less than 1 in this case. If  $w_1$  takes value closer to 1 in this case, then it is expected that the better testimator is obtained since much higher weight is given to  $\bar{X}_1$  when initial estimate is suspected. According to this principle, two alternative weighing factors and testimators are introduced as follows:

$$\text{i) } w_2 = \begin{cases} (n_1\theta_0 - Z_1) / (n_1\theta_0 - k_1), & \text{if } k_1 < Z_1 < n_1\theta_0, \\ (Z_1 - n_1\theta_0) / (k_2 - n_1\theta_0), & \text{if } n_1\theta_0 < Z_1 < k_2. \end{cases} \quad \text{and}$$

$$\hat{\theta}_2 = w_2 \bar{X}_1 + (1 - w_2)\theta_0, \quad \text{if } H_0 \text{ is accepted.}$$

Weighing factor  $w_2$  is linear functions of  $Z_1$  as  $w_1$ , and slopes of  $w_2$  are steeper than those of  $w_1$ . Weighing factor  $w_2$  is chosen since it gives weight to  $\bar{X}_1$  which is close to maximum value 1 when  $Z_1$  is within acceptance region but close to  $k_1$  or  $k_2$ . Hence, in this case,  $w_2$  gives much higher weight to  $\bar{X}_1$  than  $w_1$  does which is a desirable property. On the other hand, compared with  $w_1$ ,  $w_2$  also gives higher weight to  $\bar{X}_1$  and relatively lower one to  $\theta_0$  when  $Z_1$  is close to  $n_1\theta_0$ (that is, when initial estimate  $\theta_0$  contains valuable information), which is a undesirable property.

$$\text{ii) } w_3 = \begin{cases} (Z_1 - n_1\theta_0)^2 / (n_1\theta_0 - k_1)^2, & \text{if } k_1 < Z_1 < n_1\theta_0, \\ (Z_1 - n_1\theta_0)^2 / (k_2 - n_1\theta_0)^2, & \text{if } n_1\theta_0 < Z_1 < k_2, \end{cases} \quad \text{and}$$

$$\hat{\theta}_3 = w_3 \bar{X}_1 + (1 - w_3) \theta_0, \quad \text{if } H_0 \text{ is accepted.}$$

Weighing factor  $w_3$  is quadratic functions of  $Z_1$ . It is tried for the same reason as  $w_2$ . That is, it gives much higher weight (upto maximum 1) to  $\bar{X}_1$  than  $w_1$  does when  $H_0$  is accepted but an initial estimate is somewhat suspected. Furthermore,  $w_3$  gives lower weight (than  $w_1$ ) to  $\bar{X}_1$  and higher one (than  $w_1$ ) to  $\theta_0$  when  $Z_1$  takes value close to  $n_1 \theta_0$ , since it is quadratic functions. The latter one is a desirable property that  $w_2$  does not have.

#### 4. Comparison of testimators

Two testimators  $\hat{\theta}_2$  and  $\hat{\theta}_3$  defined with changed weighing factors  $w_2$  and  $w_3$  are introduced in previous section. In this section, their properties are compared with those of  $\hat{\theta}_1$  by means of MSE.

MSE formulae of  $\hat{\theta}_2$  and  $\hat{\theta}_3$  are derived through tedious but straightforward calculation, and are given in the below:

$$\text{i) } \text{MSE}(\hat{\theta}_2 | \theta) = E(\hat{\theta}_2^2) - 2\theta \cdot E(\hat{\theta}_2) + \theta^2.$$

The expressions for  $E(\hat{\theta}_2)$  and  $E(\hat{\theta}_2^2)$  are,

$$E(\hat{\theta}_2) = b_{20}(\theta) + \theta \cdot b_{21}(\theta) + \theta^2 \cdot b_{22}(\theta), \quad \text{where}$$

$$b_{20}(\theta) = \theta_0 \cdot \gamma(n_1, k_1 / \theta, k_2 / \theta) + [n_1 \theta_0^2 / (k_2 - n_1 \theta_0)] \cdot \gamma(n_1, n_1 \theta_0 / \theta, k_2 / \theta) \\ - [n_1 \theta_0^2 / (n_1 \theta_0 - k_1)] \cdot \gamma(n_1, k_1 / \theta, n_1 \theta_0 / \theta),$$

$$b_{21}(\theta) = 1 - [n_2 / (n_1 + n_2)] \cdot \gamma(n_1, k_1 / \theta, k_2 / \theta) - [n_1 / (n_1 + n_2)] \gamma(n_1 + 1, k_1 / \theta, k_2 / \theta) \\ - [2 n_1 \theta_0 / (k_2 - n_1 \theta_0)] \cdot \gamma(n_1 + 1, n_1 \theta_0 / \theta, k_2 / \theta) \\ + [2 n_1 \theta_0 / (n_1 \theta_0 - k_1)] \cdot \gamma(n_1 + 1, k_1 / \theta, n_1 \theta_0 / \theta),$$

$$b_{22}(\theta) = [(n_1 + 1) / (k_2 - n_1 \theta_0)] \cdot \gamma(n_1 + 2, n_1 \theta_0 / \theta, k_2 / \theta) \\ - [(n_1 + 1) / (n_1 \theta_0 - k_1)] \cdot \gamma(n_1 + 2, k_1 / \theta, n_1 \theta_0 / \theta),$$

where  $\gamma(n, a, b) = \gamma(n, b) - \gamma(n, a)$ . And,

$$E(\hat{\theta}_2^2) = \theta^2 + [\theta^2 / (n_1 + n_2)] + \sum_{j=0}^2 c_j^* \cdot \gamma(n_1 + j, k_1 / \theta, k_2 / \theta)$$

$$+ \sum_{j=0}^4 d_{2j}^* \cdot \gamma(n_1 + j, k_1/\theta, n_1\theta_0/\theta) + \sum_{j=0}^4 e_{2j}^* \cdot \gamma(n_1 + j, n_1\theta_0/\theta, k_2/\theta),$$

where

$$\begin{aligned} c_j^* &= n_1(n_1 + 1) \cdots (n_1 + j - 1) \theta^j c_j \quad j = 0, 1, 2, \\ d_{2j}^* &= n_1(n_1 + 1) \cdots (n_1 + j - 1) \theta^j d_{2j} \quad j = 0, 1, \dots, 4, \quad \text{and} \\ e_{2j}^* &= n_1(n_1 + 1) \cdots (n_1 + j - 1) \theta^j e_{2j} \quad j = 0, 1, \dots, 4. \end{aligned}$$

Further,

$$\begin{aligned} c_0 &= [(n_1 + n_2)^2 \theta_0^2 - n_2 \theta^2 (n_2 + 1)] / (n_1 + n_2)^2, \quad c_1 = -2 n_2 \theta / (n_1 + n_2)^2, \\ c_2 &= -1 / (n_1 + n_2)^2, \\ d_{20} &= [n_1^4 \theta_0^4 - 2 n_1^3 \theta_0^3 (n_1 \theta_0 - k_1)] / n_1^2 (n_1 \theta_0 - k_1)^2, \\ d_{21} &= [-4 n_1^3 \theta_0^3 + 4 n_1^2 \theta_0^2 (n_1 \theta_0 - k_1)] / n_1^2 (n_1 \theta_0 - k_1)^2, \\ d_{22} &= [6 n_1^2 \theta_0^2 - 2 n_1 \theta_0 (n_1 \theta_0 - k_1)] / n_1^2 (n_1 \theta_0 - k_1)^2, \\ d_{23} &= -4 n_1 \theta_0 / n_1^2 (n_1 \theta_0 - k_1)^2, \quad d_{24} = 1 / n_1^2 (n_1 \theta_0 - k_1)^2, \\ e_{20} &= [n_1^4 \theta_0^4 + 2 n_1^3 \theta_0^3 (k_2 - n_1 \theta_0)] / n_1^2 (k_2 - n_1 \theta_0)^2, \\ e_{21} &= [-4 n_1^3 \theta_0^3 - 4 n_1^2 \theta_0^2 (k_2 - n_1 \theta_0)] / n_1^2 (k_2 - n_1 \theta_0)^2, \\ e_{22} &= [6 n_1^2 \theta_0^2 + 2 n_1 \theta_0 (k_2 - n_1 \theta_0)] / n_1^2 (k_2 - n_1 \theta_0)^2, \\ e_{23} &= -4 n_1 \theta_0 / n_1^2 (k_2 - n_1 \theta_0)^2, \quad \text{and} \quad e_{24} = 1 / n_1^2 (k_2 - n_1 \theta_0)^2. \end{aligned}$$

$$\text{ii) } \text{MSE}(\widehat{\theta}_3 | \theta) = E(\widehat{\theta}_3^2) - 2\theta \cdot E(\widehat{\theta}_3) + \theta^2.$$

The expressions for  $E(\widehat{\theta}_3)$  and  $E(\widehat{\theta}_3^2)$  are,

$$\begin{aligned} E(\widehat{\theta}_3) &= b_{30}(\theta) + \theta b_{31}(\theta) + \theta^2 b_{32}(\theta) + \theta^3 b_{33}(\theta), \quad \text{where} \\ b_{30}(\theta) &= \theta_0 \cdot \gamma(n_1, k_1/\theta, k_2/\theta) - [n_1^2 \theta_0^3 / (n_1 \theta_0 - k_1)^2] \cdot \gamma(n_1, k_1/\theta, n_1 \theta_0/\theta) \\ &\quad - [n_1^2 \theta_0^3 / (k_2 - n_1 \theta_0)^2] \cdot \gamma(n_1, n_1 \theta_0/\theta, k_2/\theta), \\ b_{31}(\theta) &= 1 - [n_2 / (n_1 + n_2)] \cdot \gamma(n_1, k_1/\theta, k_2/\theta) - [n_1 / (n_1 + n_2)] \gamma(n_1 + 1, k_1/\theta, k_2/\theta) \\ &\quad + [3 n_1^2 \theta_0^2 / (n_1 \theta_0 - k_1)^2] \cdot \gamma(n_1 + 1, k_1/\theta, n_1 \theta_0/\theta) \\ &\quad + [3 n_1^2 \theta_0^2 / (k_2 - n_1 \theta_0)^2] \cdot \gamma(n_1 + 1, n_1 \theta_0/\theta, k_2/\theta), \\ b_{32}(\theta) &= - [3 \theta_0 n_1 (n_1 + 1) / (n_1 \theta_0 - k_1)^2] \cdot \gamma(n_1 + 2, k_1/\theta, n_1 \theta_0/\theta) \end{aligned}$$

$$\begin{aligned}
 & - [3 \theta_0 n_1 (n_1 + 1) / (k_2 - n_1 \theta_0)^2] \cdot \gamma(n_1 + 2, n_1 \theta_0 / \theta, k_2 / \theta), \\
 b_{33}(\theta) = & [ (n_1 + 1)(n_1 + 2) / (n_1 \theta_0 - k_1)^2 ] \cdot \gamma(n_1 + 3, k_1 / \theta, n_1 \theta_0 / \theta) \\
 & + [ (n_1 + 1)(n_1 + 2) / (k_2 - n_1 \theta_0)^2 ] \cdot \gamma(n_1 + 3, n_1 \theta_0 / \theta, k_2 / \theta),
 \end{aligned}$$

and

$$\begin{aligned}
 E(\widehat{\theta}_3^2) = & \theta^2 + [\theta^2 / (n_1 + n_2)] + \sum_{j=0}^2 c_j^* \cdot \gamma(n_1 + j, k_1 / \theta, k_2 / \theta) \\
 & + \sum_{j=0}^6 d_{3j}^* \cdot \gamma(n_1 + j, k_1 / \theta, n_1 \theta_0 / \theta) + \sum_{j=0}^6 e_{3j}^* \cdot \gamma(n_1 + j, n_1 \theta_0 / \theta, k_2 / \theta),
 \end{aligned}$$

where

$$\begin{aligned}
 d_{3j}^* = & n_1(n_1 + 1) \cdots (n_1 + j - 1) \theta^j d_{3j}, \quad j = 0, 1, \dots, 6, \quad \text{and} \\
 e_{3j}^* = & n_1(n_1 + 1) \cdots (n_1 + j - 1) \theta^j e_{3j}, \quad j = 0, 1, \dots, 6.
 \end{aligned}$$

Further,

$$\begin{aligned}
 d_{30} = & [n_1^6 \theta_0^6 - 2 n_1^4 \theta_0^4 (n_1 \theta_0 - k_1)^2] / n_1^2 (n_1 \theta_0 - k_1)^4, \\
 d_{31} = & [-6 n_1^5 \theta_0^5 + 6 n_1^3 \theta_0^3 (n_1 \theta_0 - k_1)^2] / n_1^2 (n_1 \theta_0 - k_1)^4, \\
 d_{32} = & [15 n_1^4 \theta_0^4 - 6 n_1^2 \theta_0^2 (n_1 \theta_0 - k_1)^2] / n_1^2 (n_1 \theta_0 - k_1)^4, \\
 d_{33} = & [-20 n_1^3 \theta_0^3 + 2 n_1 \theta_0 (n_1 \theta_0 - k_1)^2] / n_1^2 (n_1 \theta_0 - k_1)^4, \\
 d_{34} = & 15 n_1^2 \theta_0^2 / n_1^2 (n_1 \theta_0 - k_1)^4, \quad d_{35} = -6 n_1 \theta_0 / n_1^2 (n_1 \theta_0 - k_1)^4, \\
 d_{36} = & 1 / n_1^2 (n_1 \theta_0 - k_1)^4, \\
 e_{30} = & [n_1^6 \theta_0^6 - 2 n_1^4 \theta_0^4 (k_2 - n_1 \theta_0)^2] / n_1^2 (k_2 - n_1 \theta_0)^4, \\
 e_{31} = & [-6 n_1^5 \theta_0^5 + 6 n_1^3 \theta_0^3 (k_2 - n_1 \theta_0)^2] / n_1^2 (k_2 - n_1 \theta_0)^4, \\
 e_{32} = & [15 n_1^4 \theta_0^4 - 6 n_1^2 \theta_0^2 (k_2 - n_1 \theta_0)^2] / n_1^2 (k_2 - n_1 \theta_0)^4, \\
 e_{33} = & [-20 n_1^3 \theta_0^3 + 2 n_1 \theta_0 (k_2 - n_1 \theta_0)^2] / n_1^2 (k_2 - n_1 \theta_0)^4, \quad e_{34} = 15 n_1^2 \theta_0^2 / n_1^2 (k_2 - n_1 \theta_0)^4, \\
 e_{35} = & -6 n_1 \theta_0 / n_1^2 (k_2 - n_1 \theta_0)^4, \quad \text{and} \quad e_{36} = 1 / n_1^2 (k_2 - n_1 \theta_0)^4.
 \end{aligned}$$

To compare the properties of three estimators with different weighing factors, their MSE are calculated for various values of  $n_1(10, 20, 30 \text{ and } 50)$  and  $n_2(10 \sim 60(10))$ . Three true mean values ( $\theta = 1.00, 0.90, 0.80$ ) are used with fixed initial estimate  $\theta_0 = 1$ . Two significance levels (0.05, 0.10) are included. The values of  $k_1$  and  $k_2$  are obtained from (2.1).

Table 1 contains relative efficiency (to usual single sample mean) calculation results. From these tables, the following phenomena are found for each value of  $\theta$ .

(1) When  $\theta_0$  is very close to  $\theta$  (that is, when  $\theta \approx \theta_0 = 1.0$ ), it follows that

$$eff_{\bar{X}, \hat{\theta}_1}(\theta) > eff_{\bar{X}, \hat{\theta}_2}(\theta) > eff_{\bar{X}, \hat{\theta}_3}(\theta)$$

for all cases. That is,  $\hat{\theta}_1$  is the best and  $\hat{\theta}_3$  is the second, regardless of  $n_1, n_2$  and  $\alpha$  values.

(2) When  $\theta_0$  is somewhat close to  $\theta$  (that is, when  $\theta \approx 0.90$ ), it also follows that

$$eff_{\bar{X}, \hat{\theta}_1}(\theta) > eff_{\bar{X}, \hat{\theta}_3}(\theta) > eff_{\bar{X}, \hat{\theta}_2}(\theta)$$

for all cases. But, the differences among efficiencies are much less than those of (1).

(3) When  $\theta_0$  is relatively far from  $\theta$  (that is, when  $\theta \approx 0.80$ ) and  $n_1 = 10$ , it also follows that

$$eff_{\bar{X}, \hat{\theta}_1}(\theta) > eff_{\bar{X}, \hat{\theta}_3}(\theta) > eff_{\bar{X}, \hat{\theta}_2}(\theta).$$

But for other larger values of  $n_1$  ( $= 20, 30, 50$ ), we have the following different result.

$$eff_{\bar{X}, \hat{\theta}_2}(\theta) > eff_{\bar{X}, \hat{\theta}_3}(\theta) > eff_{\bar{X}, \hat{\theta}_1}(\theta).$$

(4) The differences among efficiencies when  $\alpha = 0.10$  are reduced a little bit when compared with those of  $\alpha = 0.05$  in all cases included in Table 1.

Table 1 contains the relative efficiencies only for  $\theta_0 = 1$ . Finally, one theorem that extends the applicability of above results (1) ~ (4) for various combinations of  $\theta$  and  $\theta_0$  is provided.

**Theorem.**  $Eff_{\bar{X}, \hat{\theta}_i}(\theta)$ ,  $i = 1, 2, 3$ , are invariant to the ratio  $\theta/\theta_0$ .

**Proof.** It is proved in Moon(1998) for  $\hat{\theta}_1$ . Proofs for  $\hat{\theta}_2$  and  $\hat{\theta}_3$  can be made similarly.

## 5. Concluding Remarks

Two testimators with changed weighing factors are introduced with the hope that they are superior to existing Adke, Waikar and Schuurmann's testimator in MSE sense. Their MSE formulae are derived. And their MSE values are calculated for various values of  $\theta, n_1, n_2$  and  $\alpha$  with fixed  $\theta_0 = 1$ . From the comparison of efficiencies (relative to usual single sample mean) of three testimators, the following conclusions are made. Note that with fixed

$n_1$ ,  $n_2$  and  $\alpha$ , conclusions hold for various combinations of  $\theta$  and  $\theta_0$  values if  $\theta/\theta_0$  is the same since the relative efficiencies are invariant to  $\theta/\theta_0$ .

(1) When we have a valuable initial estimate (that is, when  $\theta_0$  is relatively close to  $\theta$ ), the behavior of  $\hat{\theta}_1$  is the best and that of  $\hat{\theta}_3$  is the second in all cases included in Table 1. On the other hand, if we have an initial information that is relatively far from  $\theta$  and relatively larger  $n_1 (= 20, 30, 50)$ , then the behavior of  $\hat{\theta}_2$  is the best and that of  $\hat{\theta}_3$  is the second. But for a relatively smaller  $n_1 (= 10)$  in this case, the result is the same as that of a valuable initial estimate case.

(2) As the significance level  $\alpha$  gets larger, the differences among three efficiencies decrease.

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Table 1. Efficiency of testimators w.r.t.  $\bar{X}$  when  $\theta_0 = 1$ .

$n_1 = 10, \alpha = 0.05, k_1 = 4.97890, k_2 = 17.61340.$

$\vartheta$	$\hat{\vartheta}_1$			$\hat{\vartheta}_2$			$\hat{\vartheta}_3$		
$n_2 \backslash \theta$	1.00	0.90	0.80	1.00	0.90	0.80	1.00	0.90	0.80
10	5.58120	3.97365	1.75482	2.28459	2.09691	1.55984	3.29221	2.81277	1.65386
20	6.93123	4.40497	1.73743	2.40892	2.14881	1.53054	3.63932	2.96812	1.62986
30	7.49062	4.48089	1.65816	2.40002	2.10682	1.45465	3.70459	2.94885	1.55212
40	7.67743	4.42528	1.56767	2.34895	2.03843	1.37204	3.66817	2.87382	1.46565
50	7.68252	4.31703	1.47936	2.28336	1.96326	1.29285	3.59113	2.78022	1.38199
60	7.59445	4.18817	1.39714	2.21350	1.88817	1.21975	3.49764	2.68186	1.30452

$n_1 = 10, \alpha = 0.10, k_1 = 5.62920, k_2 = 16.19890.$

$\vartheta$	$\hat{\vartheta}_1$			$\hat{\vartheta}_2$			$\hat{\vartheta}_3$		
$n_2 \backslash \theta$	1.00	0.90	0.80	1.00	0.90	0.80	1.00	0.90	0.80
10	4.36230	3.22885	1.56084	2.17364	1.95639	1.44522	2.85548	2.44627	1.50993
20	5.73503	3.67568	1.51590	2.34644	2.02121	1.39148	3.26447	2.62057	1.46090
30	6.37156	3.74433	1.41126	2.32701	1.95687	1.28983	3.33448	2.58365	1.35745
40	6.60106	3.66597	1.30238	2.24585	1.85879	1.18729	3.27401	2.47996	1.25128
50	6.61431	3.53131	1.20227	2.14619	1.75549	1.09418	3.16333	2.35803	1.15425
60	6.51405	3.37812	1.11331	2.04393	1.65635	1.01202	3.03540	2.23525	1.06825

$n_1 = 20, \alpha = 0.05, k_1 = 12.43910, k_2 = 30.136710.$

$\vartheta$	$\hat{\vartheta}_1$			$\hat{\vartheta}_2$			$\hat{\vartheta}_3$		
$n_2 \backslash \theta$	1.00	0.90	0.80	1.00	0.90	0.80	1.00	0.90	0.80
10	4.42137	2.69535	1.03681	2.08104	1.79785	1.16621	2.94496	2.18563	1.08486
20	5.66076	3.01932	1.03772	2.28721	1.91194	1.17815	3.41508	2.37628	1.08956
30	6.51292	3.17131	1.01049	2.37917	1.94633	1.15354	3.67024	2.45050	1.06314
40	7.07846	3.23121	0.97343	2.41411	1.94342	1.11519	3.80334	2.46725	1.02547
50	7.44816	3.24015	0.93370	2.41868	1.92185	1.07228	3.86490	2.45380	0.98446
60	7.67877	3.21956	0.89426	2.40619	1.89065	1.02882	3.88441	2.42383	0.94352

$n_1 = 20, \alpha = 0.10, k_1 = 13.49340, k_2 = 28.32260.$

$\vartheta$	$\hat{\vartheta}_1$			$\hat{\vartheta}_2$			$\hat{\vartheta}_3$		
$n_2 \backslash \theta$	1.00	0.90	0.80	1.00	0.90	0.80	1.00	0.90	0.80
10	3.29542	2.22126	0.98193	1.90670	1.65080	1.10661	2.49020	1.90759	1.02180
20	4.39450	2.53744	0.96297	2.17830	1.78714	1.09804	3.02722	2.11501	1.00590
30	5.22068	2.68568	0.91964	2.30636	1.82574	1.05633	3.34452	2.19414	0.96282
40	5.81062	2.73832	0.87008	2.35472	1.81648	1.00437	3.51814	2.20570	0.91237
50	6.21430	2.73596	0.82089	2.35842	1.78342	0.95098	3.60069	2.18167	0.86175
60	6.47978	2.70271	0.77450	2.33671	1.73869	0.89963	3.62563	2.13831	0.81372

Table 1. (Continued)

$n_1 = 30, \alpha = 0.05, k_1 = 20.482370, k_2 = 42.089230.$

$\hat{\theta}$	$\hat{\theta}_1$			$\hat{\theta}_2$			$\hat{\theta}_3$		
$n_2 \backslash \theta$	1.00	0.90	0.80	1.00	0.90	0.80	1.00	0.90	0.80
10	3.91843	2.09719	0.77115	1.96827	1.60360	0.96727	2.63980	1.79571	0.84800
20	4.90054	2.30635	0.76587	2.16902	1.71103	0.97556	3.04239	1.93704	0.84868
30	5.68743	2.42675	0.74741	2.28859	1.76353	0.96195	3.30292	2.00925	0.83489
40	6.30122	2.49199	0.72346	2.35858	1.78441	0.93796	3.48466	2.05436	0.80801
50	6.77060	2.52207	0.69761	2.39753	1.78654	0.90936	3.61639	2.06028	0.78131
60	7.12868	2.52943	0.67154	2.41642	1.77714	0.87908	3.66571	2.04666	0.75799

$n_1 = 30, \alpha = 0.10, k_1 = 21.848860, k_2 = 39.962770.$

$\hat{\theta}$	$\hat{\theta}_1$			$\hat{\theta}_2$			$\hat{\theta}_3$		
$n_2 \backslash \theta$	1.00	0.90	0.80	1.00	0.90	0.80	1.00	0.90	0.80
10	2.87105	1.77519	0.76759	1.76517	1.47143	0.94135	2.16127	1.59494	0.85581
20	3.69572	1.97405	0.74563	2.01663	1.59092	0.92854	2.58625	1.73306	0.83429
30	4.40088	2.08766	0.71419	2.17614	1.64832	0.89912	2.85573	1.82246	0.80171
40	4.98469	2.14591	0.68015	2.27281	1.66842	0.86319	3.04778	1.86207	0.76394
50	5.45400	2.16801	0.64645	2.32710	1.66582	0.82550	3.17554	1.86955	0.72573
60	5.82847	2.16681	0.61434	2.35270	1.64935	0.78835	3.30310	1.84578	0.69823

$n_1 = 50, \alpha = 0.05, k_1 = 37.37230, k_2 = 65.19630.$

$\hat{\theta}$	$\hat{\theta}_1$			$\hat{\theta}_2$			$\hat{\theta}_3$		
$n_2 \backslash \theta$	1.00	0.90	0.80	1.00	0.90	0.80	1.00	0.90	0.80
10	3.47508	1.51087	0.56470	1.85906	1.35148	0.79043	2.45108	1.39652	0.66354
20	4.13709	1.60937	0.55172	2.01800	1.42784	0.78631	2.74128	1.47999	0.65426
30	4.73204	1.67586	0.53566	2.13431	1.47938	0.76877	2.98256	1.53629	0.63529
40	5.25516	1.71920	0.51832	2.21561	1.50024	0.75770	3.13803	1.56319	0.62317
50	5.70766	1.74566	0.50068	2.28370	1.52178	0.73491	3.29372	1.58506	0.60259
60	6.10021	1.76025	0.48330	2.32526	1.53354	0.70807	3.38022	1.60140	0.58111

$n_1 = 50, \alpha = 0.10, k_1 = 39.23560, k_2 = 62.57060.$

$\hat{\theta}$	$\hat{\theta}_1$			$\hat{\theta}_2$			$\hat{\theta}_3$		
$n_2 \backslash \theta$	1.00	0.90	0.80	1.00	0.90	0.80	1.00	0.90	0.80
10	2.52245	1.34513	0.60731	1.62666	1.25586	0.80847	1.94935	1.28190	0.70582
20	3.05854	1.43535	0.58123	1.81830	1.33654	0.78318	2.23770	1.36867	0.67879
30	3.55932	1.49438	0.55493	1.96454	1.38333	0.75331	2.47787	1.41593	0.64901
40	4.01834	1.53079	0.52952	2.09599	1.42063	0.72834	2.70644	1.45765	0.62558
50	4.43495	1.55104	0.50547	2.17530	1.43001	0.70091	2.85525	1.47020	0.60041
60	4.80761	1.55915	0.48288	2.24331	1.43018	0.67324	2.97499	1.45926	0.57140