

# Shrinkage Testimator in Gamma Type-II Censored Data under LINEX Loss Function

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## ABSTRACT

Prakash and Singh presented the shrinkage estimators under the invariant version of LINEX loss function for the scale parameter of an exponential distribution in presence Type-II censored data. In this paper, we extend this approach to gamma distribution, as Prakash and Singh's paper is a special case of this paper. In fact, some shrinkage estimators for the scale parameter of a gamma distribution, when Type-II censored data are available, have been suggested under the LINEX loss function assuming the shape parameter is to be known. The comparisons of the proposed estimators have been made with improved estimator. All these estimators are compared empirically using Monte Carlo simulation.

**Keywords:** Gamma Distribution; Shrinkage Estimator and Factor; Asymmetric Loss Function; Level of Significance; Testimation; Monte-Carlo Simulation

## 1. Introduction

In life-testing research, the most widely used life distribution is the Gamma with probability density function for any random variable  $x$ :

$$f(x; \lambda, \theta) = \frac{x^{\lambda-1}}{\Gamma(\lambda)\theta^\lambda} e^{-\frac{x}{\theta}}; \quad x \geq 0, \lambda, \theta \geq 0. \quad (1.1)$$

Let  $x_1, x_2, \dots, x_n$  be the random samples of size  $n$  taken from the Gamma distribution. The parameter  $\lambda$  and  $\theta$  are called the shape and scale parameter, respectively. It is crucial to have in-depth study of the (Classic and Bayes) estimate of the scale parameter of Gamma distribution because, in several cases, the distribution of the minimal sufficient statistics is Gamma (see Parsian and Kirmani [1]). Pazira and Shadrokh [2] derived Bayes estimators of the scale parameter of gamma distribution on the two asymmetric loss function LINEX and Precautionary by using several prior distributions and then compared the efficiency of all estimates. In the present paper, concentration is on the gamma distribution.

Ferguson [3], Zellner and Geisel [4], Aitchison and Dunsmore [5], Varian [6], and Berger [7] indicated to insufficient to symmetric loss function and just Varian [6] suggested asymmetric linear loss function. This loss function was widely used by several authors; among of them were Basu and Ebrahimi [8], Pandey [9], Soliman [10], and Prakash and Singh [11]. Following Basu and

Ebrahimi [8], the invariant form of the LINEX loss function (ILL) for any parameter  $\theta$  is defined as

$$L(\Delta) = \{e^{c\Delta} - c\Delta - 1\}; \quad c \neq 0, \Delta = \left( \frac{\hat{\theta}}{\theta} - 1 \right), \quad (1.2)$$

where  $c$  is the shape parameter and  $\hat{\theta}$  is any estimate of the parameter  $\theta$ .

The LINEX loss function is convex and the shape of this loss function is determined by the value of  $c$ . The negative (positive) value of  $c$  gives more weight to overestimation (underestimation) and its magnitude reflects the degree of asymmetry. It is seen that, for  $c = 1$ , the function is quite asymmetric with overestimation being costlier than underestimation. If  $c < 0$ , it rises almost exponentially when the estimation error  $(\hat{\theta} - \theta) < 0$  and almost linearly when  $(\hat{\theta} - \theta) > 0$ . For small values of  $|c|$ , the LINEX loss function is almost symmetric and not far from squared error loss function.

Pandey [9], Parsian and Farsipour [12], Singh, Gupta, and Upadhyay [13], Misra and Meulen [14], Ahmadi, Doostparast, and Parsian [15], Xiao, Takada, and Shi [16], Singh, Prakash, and Singh [17] and others have used the LINEX loss function in the various estimation and prediction problems.

In life-testing, fatigue failures and other kinds of destructive test situations, the observations usually occurred in an ordered manner such a way that the weakest items

failed first and then the second one and so on. Let us suppose that  $n$  items are put on life test and terminate the experiment when  $r$  ( $< n$ ) items have failed. If  $X_{(1)}, X_{(2)}, \dots, X_{(r)}$  denote the first  $r$  observations having a common density function as given in (1.1) then the joint probability density function is given by

$$\begin{aligned} & f(x_{(1)}, x_{(2)}, \dots, x_{(r)} | \theta) \\ &= \frac{n!}{(n-r)!} \left(\frac{1}{\theta}\right)^{r\lambda} \left(\frac{1}{\Gamma(\lambda)}\right)^r \left(\prod_{i=1}^r x_{(i)}^{\lambda-1}\right) e^{-\frac{T_r}{\theta}} \end{aligned} \quad (1.3)$$

where

$$T_r = \sum_{i=1}^r x_{(i)} + (n-r)x_{(r)} \quad (1.4)$$

$T_r$  is a complete sufficient statistic of  $\theta$  and distributed as gamma distribution with parameters  $(r\lambda, r\theta)$ . The maximum likelihood estimator (MLE) of  $\theta$  is given by

$$\hat{\theta} = \frac{T_r}{r\lambda} \sim \frac{\theta}{2r\lambda} \chi^2_{(2r\lambda)} \quad (1.5)$$

and can easily show that  $\hat{\theta}$  is the minimum variance unbiased estimator (MVUE) of  $\theta$ .

Roa and Srivastava [18] considered a class for the total test time as

$$Y = d \frac{T_r}{r\lambda},$$

and found the value of the constant

$$d = \frac{r\lambda}{c} \left(1 - e^{-\frac{c}{r\lambda+1}}\right) = d_1$$

(say) which minimizes the risk of  $Y$  under the ILL. The minimum risk estimator is

$$Y_1 = d_1 \frac{T_r}{r\lambda}$$

with the minimum risk under ILL

$$Risk(Y_1) = c + (r\lambda + 1) \left(e^{-\frac{c}{r\lambda+1}} - 1\right), \quad (1.6)$$

for  $\lambda = 1$ , see Prakash and Singh [11].

In the present paper, some shrinkage estimators for the scale parameter of a gamma distribution, when Type-II censored data are available, have been suggested under the ILL loss function assuming the shape parameter is to be known.

## 2. Shrinkage Testimators and their Properties

Following Thompson [19], the shrinkage estimator for the parameter  $\theta$  is given by

$$\hat{Y} = \theta_0 + k \left( \frac{T_r}{r\lambda} - \theta_0 \right), \quad 0 \leq k \leq 1. \quad (2.1)$$

The value of the shrinkage factor  $k$  near to the zero implies strong belief in the guess value  $\theta_0$  and near to one implies a strong belief in the sample values. Several researchers have studied the performance of the shrinkage estimators and found that the shrinkage estimator performs better with respect to any usual estimator when the guess value  $\theta_0$  is close to the parameter  $\theta$ . This suggests that we may test the hypothesis  $H_0 : \theta = \theta_0$  against  $H_1 : \theta \neq \theta_0$ . A test statistic

$$\frac{2T_r}{\theta_0} \sim \chi^2_{(2r\lambda)}$$

is available for testing the hypothesis  $H_0$ .

The loss for estimator  $\hat{Y}$  under the ILL is defined as

$$L(\hat{Y}) = \{e^{c\Delta} - c\Delta - 1\},$$

where

$$c\Delta = c \{ \delta - 1 + k(T_r / (r\lambda\theta) - \delta) \}$$

and

$$\delta = \frac{\theta_0}{\theta}.$$

The risk of the proposed shrinkage estimator  $\hat{Y}$  under the ILL is given by

$$Risk(\hat{Y}) = e^{c(\delta-1)} e^{-ck\delta} \left(1 - \frac{ck}{r\lambda}\right)^{-r\lambda} - c(\delta-1)(1-k)-1 \quad (2.2)$$

The value of  $k = k_{\min}$  (say), which minimizes the risk  $Risk(\hat{Y})$  is thus obtained by solving the given Equation

$$(1-\delta)e^{c(1-\delta)} = e^{-ck\delta} \left(1 - \frac{ck}{r\lambda}\right)^{-(r\lambda+1)} \left\{1 - \delta \left(1 - \frac{ck}{r\lambda}\right)\right\}. \quad (2.3)$$

The value of  $k_{\min}$  depends upon the unknown parameter  $\theta$ . Hence, an estimate  $\hat{k}$  of  $k_{\min}$  is obtained by replacing the parameter  $\theta$  to its minimum variance unbiased estimator. Based on this, the proposed shrinkage estimator for the scale parameter  $\theta$  is defined as

$$\hat{\theta}_1 = d_1 \frac{T_r}{r\lambda} + \left( (1-\hat{k})\theta_0 + (\hat{k}-d_1) \frac{T_r}{r\lambda} \right) I_{(t_1 \leq T_r \leq t_2)}, \quad (2.4)$$

where  $I_{(A)}$  denotes the indicator of  $A$ ,  $t_1 = m_1\theta_0/2$  and  $t_2 = m_2\theta_0/2$ . Here  $m_1$  and  $m_2$  are the values of the lower and upper

$$100 \frac{\alpha}{2} \%$$

points of the chi-square distribution with  $2r\lambda$  degrees of freedom. The risk under the ILL for the shrinkage estimator  $\hat{\theta}_1$  is given by

$$\begin{aligned} Risk(\hat{\theta}_1) &= \left\{ e^{c(\delta-1)} G(w_1, w_2, e^f) - e^{-c} G(w_1, w_2, f_0) \right. \\ &\quad \left. - G(w_1, w_2, f) - c\delta G(w_1, w_2, 1) + G(w_1, w_2, f_0) \right. \\ &\quad \left. + c + (1+r\lambda) \left( \exp\left(-\frac{c}{r\lambda+1}\right) - 1 \right) \right\}, \end{aligned} \quad (2.5)$$

where

$$\begin{aligned} w_1 &= \delta m_1/2, \quad w_2 = \delta m_2/2, \quad f = c\hat{k}(w/r\lambda - \delta), \\ f_0 &= (wcd_1/r\lambda), \quad G(u, v, y) = \frac{1}{\Gamma(r\lambda)} \int_u^y yw^{r\lambda-1} e^{-w} dw \end{aligned}$$

and  $y$  may be a function of  $w$ . For  $\lambda=1$ , see Prakash and Singh [11].

Waikar, Schuurmann, and Raghunathan [20] has suggested an idea of selecting the shrinkage factor which is the function of the test statistic *i.e.*, under  $H_0: \theta = \theta_0$

$$\begin{aligned} m_1 &\leq \frac{2T_r}{\theta_0} \leq m_2 \\ \Leftrightarrow 0 \leq k_1 \text{ (say)} &= \frac{1}{m_2 - m_1} \left( \frac{2T_r}{\theta_0} - m_1 \right) \leq 1. \end{aligned}$$

Therefore, the proposed shrinkage estimator based on  $k_1$  is given by

$$\hat{\theta}_2 = d_1 \frac{T_r}{r\lambda} + \left( (1-k_1)\theta_0 + (k_1-d_1) \frac{T_r}{r\lambda} \right) I_{(t_1 \leq T_r \leq t_2)}. \quad (2.6)$$

The risk under the ILL for the shrinkage estimator  $\hat{\theta}_2$  is given by

$$\begin{aligned} Risk(\hat{\theta}_2) &= \left\{ e^{c(\delta-1)} G(w_1, w_2, e^{f_1}) - e^{-c} G(w_1, w_2, f_0) \right. \\ &\quad \left. - G(w_1, w_2, f_1) - c\delta G(w_1, w_2, 1) + G(w_1, w_2, f_0) \right. \\ &\quad \left. + c + (1+r\lambda) \left( \exp\left(-\frac{c}{r\lambda+1}\right) - 1 \right) \right\}, \end{aligned} \quad (2.7)$$

where

$$f_1 = \frac{c}{m_2 - m_1} \left( \frac{2w}{\delta} - m_1 \right) \left( \frac{w}{r\lambda} - \delta \right).$$

For  $\lambda=1$ , see Prakash and Singh [11].

When  $H_0: \theta = \theta_0$  is accepted,

$$m_1 \leq 2r\lambda \leq m_2 \Rightarrow m_1/2r\lambda \leq 1.$$

If one is interested in taking smaller values of the shrinkage factor, he can take  $m_1/2r\lambda \geq 1$ . The proposed shrinkage estimator is

$$\hat{\theta}_3 = d_1 \frac{T_r}{r\lambda} + \left( (1-k_2)\theta_0 + (k_2-d_1) \frac{T_r}{r\lambda} \right) I_{(t_1 \leq T_r \leq t_2)} \quad (2.8)$$

where

$$k_2 = \frac{2r\lambda}{m_2 - m_1} \left| \frac{T_r}{r\lambda\theta_0} - 1 \right|;$$

it may be possible that the value of shrinkage factor is negative so positive is taken. Adke, Waikar, and Schuurmann [21] and Pandey, Malik, and Srivastava [22] have considered this type of shrinkage factor. The risk of the shrinkage estimator  $\hat{\theta}_3$  is given by

$$\begin{aligned} Risk(\hat{\theta}_3) &= \left\{ e^{c(\delta-1)} G(w_1, w_2, e^{f_2}) - e^{-c} G(w_1, w_2, f_0) \right. \\ &\quad \left. - G(w_1, w_2, f_2) - c\delta G(w_1, w_2, 1) + G(w_1, w_2, f_0) \right. \\ &\quad \left. + c + (1+r\lambda) \left( \exp\left(-\frac{c}{r\lambda+1}\right) - 1 \right) \right\}, \end{aligned} \quad (2.9)$$

where

$$f_2 = \frac{c}{m_2 - m_1} \left| \frac{2w}{\delta} - 2r\lambda \right| \left( \frac{w}{r\lambda} - \delta \right).$$

For  $\lambda=1$ , see Prakash and Singh [11].

The minimum value of constant  $d$ ,  $d_1$  obtained for the class  $Y = dT_r/r\lambda$ , lies between zero and one. Hence, it may be a choice for the shrinkage factor. Thus, the proposed shrinkage estimator may be considered as

$$\hat{\theta}_4 = d_1 \frac{T_r}{r\lambda} + ((1-d_1)\theta_0) I_{(t_1 \leq T_r \leq t_2)} \quad (2.10)$$

The risk of the proposed shrinkage estimator  $\hat{\theta}_4$  under ILL is given by

$$\begin{aligned} Risk(\hat{\theta}_4) &= \left\{ e^{c(\delta-1)} G(w_1, w_2, e^{f_1}) \right. \\ &\quad \left. - e^{-c} G(w_1, w_2, f_0) - c\delta(1+d_1) G(w_1, w_2, 1), \right. \\ &\quad \left. + c + (1+r\lambda) \left( \exp\left(-\frac{c}{r\lambda+1}\right) - 1 \right) \right\} \end{aligned} \quad (2.11)$$

where

$$f_3 = \frac{cd_1}{m_2 - m_1} \left( \frac{w}{r\lambda} - \delta \right).$$

For  $\lambda=1$ , see Prakash and Singh [11].

### 3. Numerical Illustration

The relative efficiency for  $\hat{\theta}_i$ ;  $i=1, \dots, 4$ , with respect to the minimum risk improved estimator under the ILL is defined as

$$RE\left(\hat{\theta}_i, Y_i\right) = \frac{Risk(Y_i)}{Risk(\hat{\theta}_i)}; \quad i = 1, \dots, 4$$

The expression for the relative efficiency  $RE\left(\hat{\theta}_i, Y_i\right)$ ;  $i = 1, \dots, 4$ , is the function of  $r, c, \delta, \lambda$  and  $\alpha$ . For the

selected values of  $r = 6, 8, 10$ ;  $c = 0.25, 0.5, 1, 1.5$ ;  $\delta = 0.4(0.2)1.8$ ;  $\alpha = 0.01, 0.05, 0.1$  and  $\lambda = 0.5, 2.5$ , the relative efficiencies have been calculated and presented in **Tables 1-8**. Only positive values of  $c$  are considered because overestimation in mean life is more serious

**Table 1.**  $RE\left(\hat{\theta}_i, Y_i\right)$  when  $\lambda = 0.5$ .

$c$	$\alpha$	$\delta$							
		0.4	0.6	0.8	1.0	1.2	1.4	1.6	1.8
$r = 6$									
0.25	0.01	1.0116	1.0100	0.9997	0.9997	0.9997	0.9997	0.9997	0.9997
	0.05	1.0138	1.0110	1.0102	1.0100	1.0101	1.0102	1.0103	1.0106
	0.10	1.0158	1.0120	1.0108	1.0105	1.0106	1.0108	1.0112	1.0116
0.50	0.01	1.0463	1.0442	1.0446	1.0451	1.0449	1.0444	1.0439	1.0436
	0.05	1.0508	1.0463	1.0459	1.0464	1.0466	1.0466	1.0465	1.0468
	0.10	1.0552	1.0492	1.0481	1.0486	1.0491	1.0496	1.0500	1.0508
1.00	0.01	1.1600	1.1840	1.2050	1.2150	1.2090	1.1930	1.1780	1.1680
	0.05	1.1570	1.1790	1.2030	1.2180	1.2150	1.2000	1.1840	1.1720
	0.10	1.1610	1.1810	1.2070	1.2250	1.2240	1.2090	1.1920	1.1790
1.50	0.01	1.2890	1.3810	1.4660	1.5170	1.4760	1.3910	1.3190	1.2720
	0.05	1.2660	1.3480	1.4470	1.5160	1.4830	1.3950	1.3200	1.2700
	0.10	1.2640	1.3400	1.4440	1.5230	1.4960	1.4070	1.3270	1.2720
$r = 8$									
0.25	0.01	1.0100	0.9982	0.9978	0.9978	0.9978	0.9978	0.9978	0.9979
	0.05	1.0124	0.9991	0.9983	0.9981	0.9982	0.9983	0.9985	0.9988
	0.10	1.0144	1.0101	0.9989	0.9986	0.9987	0.9990	0.9994	1.0100
0.50	0.01	1.0383	1.0358	1.0359	1.0362	1.0361	1.0358	1.0356	1.0356
	0.05	1.0432	1.0383	1.0375	1.0379	1.0382	1.0384	1.0387	1.0393
	0.10	1.0481	1.0415	1.0400	1.0404	1.0410	1.0417	1.0426	1.0438
1.00	0.01	1.1250	1.1460	1.1630	1.1710	1.1660	1.1560	1.1460	1.1400
	0.05	1.1260	1.1450	1.1660	1.1800	1.1780	1.1660	1.1550	1.1480
	0.10	1.1330	1.1510	1.1750	1.1940	1.1930	1.1800	1.1670	1.1580
1.50	0.01	1.2210	1.2970	1.3650	1.4040	1.3740	1.3150	1.2690	1.2410
	0.05	1.2090	1.2790	1.3660	1.4280	1.4000	1.3310	1.2760	1.2420
	0.10	1.2140	1.2820	1.3810	1.4620	1.4360	1.3540	1.2890	1.2480
$r = 10$									
0.25	0.01	0.9990	0.9969	0.9966	0.9965	0.9965	0.9966	0.9966	0.9967
	0.05	1.0115	0.9980	0.9970	0.9969	0.9969	0.9971	0.9974	0.9977
	0.10	1.0138	0.9990	0.9976	0.9973	0.9975	0.9978	0.9983	0.9990
0.50	0.01	1.0331	1.0302	1.0301	1.0304	1.0303	1.0302	1.0301	1.0303
	0.05	1.0386	1.0332	1.0322	1.0325	1.0328	1.0331	1.0337	1.0346
	0.10	1.0442	1.0369	1.0350	1.0353	1.0360	1.0370	1.0382	1.0398
1.00	0.01	1.1030	1.1210	1.1360	1.1430	1.1390	1.1310	1.1250	1.1220
	0.05	1.1070	1.1240	1.1450	1.1590	1.1560	1.1460	1.1380	1.1330
	0.10	1.1170	1.1340	1.1590	1.1800	1.1790	1.1660	1.1540	1.1460
1.50	0.01	1.1770	1.2440	1.3030	1.3350	1.3110	1.2670	1.2350	1.2170
	0.05	1.1750	1.2380	1.3230	1.3840	1.3570	1.2950	1.2510	1.2240
	0.10	1.1860	1.2500	1.3570	1.4510	1.4190	1.3330	1.2710	1.2350

**Table 2.**  $RE(\hat{\theta}_2, Y_1)$  when  $\lambda = 0.5$ .

c	$\alpha$	$\delta$							
		0.4	0.6	0.8	1.0	1.2	1.4	1.6	1.8
<i>r = 6</i>									
0.25	0.01	1.0116	1.0100	0.9997	0.9996	0.9996	0.0096	0.9996	0.9996
	0.05	1.0139	1.0110	1.0102	1.0100	1.0100	1.0101	1.0103	1.0104
	0.10	1.0158	1.0120	1.0108	1.0105	1.0106	1.0108	1.0111	1.0114
0.50	0.01	1.0465	1.0441	1.0440	1.0441	1.0441	1.0435	1.0425	1.0408
	0.05	1.0510	1.0465	1.0456	1.0456	1.0457	1.0455	1.0449	1.0438
	0.10	1.0555	1.0494	1.0479	1.0479	1.0483	1.0484	1.0483	1.0476
1.00	0.01	1.1630	1.1810	1.1890	1.1910	1.1860	1.1730	1.1530	1.1270
	0.05	1.1590	1.1800	1.1950	1.1980	1.1910	1.1770	1.1560	1.1320
	0.10	1.1630	1.1830	1.2030	1.2100	1.2030	1.1860	1.1640	1.1390
1.50	0.01	1.2980	1.3640	1.3760	1.3640	1.3370	1.2900	1.2280	1.1660
	0.05	1.2730	1.3550	1.4050	1.3970	1.3510	1.2900	1.2250	1.1680
	0.10	1.2700	1.3500	1.4240	1.4330	1.3810	1.3060	1.2340	1.1750
<i>r = 8</i>									
0.25	0.01	1.0100	0.9982	0.9978	0.9977	0.9977	0.9978	0.9978	0.9978
	0.05	1.0124	0.9992	0.9983	0.9981	0.9981	0.9983	0.9985	0.9987
	0.10	1.0145	1.0102	0.9989	0.9986	0.9987	0.9989	0.9993	0.9998
0.50	0.01	1.0384	1.0357	1.0355	1.0357	1.0356	1.0353	1.0344	1.0332
	0.05	1.0434	1.0384	1.0373	1.0374	1.0376	1.0377	1.0374	1.0367
	0.10	1.0483	1.0416	1.0399	1.0399	1.0405	1.0409	1.0411	1.0410
1.00	0.01	1.1270	1.1450	1.1540	1.1570	1.1540	1.1440	1.1260	1.1050
	0.05	1.1270	1.1460	1.1620	1.1680	1.1640	1.1510	1.1330	1.1120
	0.10	1.1340	1.1530	1.1730	1.1830	1.1790	1.1640	1.1430	1.1210
1.50	0.01	1.2260	1.2920	1.3180	1.3190	1.3010	1.2580	1.1990	1.1430
	0.05	1.2130	1.2850	1.3450	1.3530	1.3200	1.2650	1.2030	1.1490
	0.10	1.2170	1.2890	1.3710	1.3970	1.3550	1.2840	1.2150	1.1580
<i>r = 10</i>									
0.25	0.01	0.9990	0.9969	0.9965	0.9965	0.9965	0.9965	0.9966	0.9966
	0.05	1.0116	0.9980	0.9970	0.9968	0.9969	0.9971	0.9973	0.9976
	0.10	1.0138	0.9990	0.9976	0.9973	0.9974	0.9978	0.9982	0.9988
0.50	0.01	1.0332	1.0302	1.0299	1.0300	1.0300	1.0297	1.0291	1.0281
	0.05	1.0387	1.0333	1.0321	1.0321	1.0324	1.0326	1.0325	1.0322
	0.10	1.0443	1.0370	1.0350	1.0350	1.0356	1.0363	1.0369	1.0371
1.00	0.01	1.1040	1.1210	1.1310	1.1340	1.1320	1.1230	1.1080	1.0895
	0.05	1.1080	1.1260	1.1430	1.1500	1.1460	1.1350	1.1180	1.0987
	0.10	1.1170	1.1360	1.1590	1.1710	1.1670	1.1520	1.1320	1.1110
1.50	0.01	1.1800	1.2430	1.2750	1.2810	1.2670	1.2280	1.1740	1.1240
	0.05	1.1770	1.2430	1.3100	1.3270	1.2990	1.2450	1.1840	1.1330
	0.10	1.1880	1.2570	1.3520	1.3930	1.3510	1.2730	1.2010	1.1460

**Table 3.**  $RE(\hat{\theta}_3, Y_1)$  when  $\lambda = 0.5$ .

c	$\alpha$	$\delta$							
		0.4	0.6	0.8	1.0	1.2	1.4	1.6	1.8
<i>r = 6</i>									
0.25	0.01	1.0116	1.0100	0.9997	0.9997	0.9997	0.9997	0.9997	0.9997
	0.05	1.0138	1.0110	1.0102	1.0100	1.0100	1.0102	1.0103	1.0105
	0.10	1.0158	1.0120	1.0108	1.0105	1.0106	1.0108	1.0112	1.0115
0.50	0.01	1.0462	1.0442	1.0444	1.0447	1.0447	1.0442	1.0433	1.0419
	0.05	1.0507	1.0464	1.0458	1.0461	1.0464	1.0464	1.0460	1.0453
	0.10	1.0552	1.0493	1.0480	1.0483	1.0489	1.0494	1.0496	1.0494
1.00	0.01	1.1590	1.1840	1.2000	1.2060	1.2020	1.1890	1.1670	1.1430
	0.05	1.1560	1.1790	1.2010	1.2110	1.2090	1.1960	1.1750	1.1530
	0.10	1.1600	1.1810	1.2050	1.2200	1.2190	1.2060	1.1850	1.1620
1.50	0.01	1.2880	1.3770	1.4380	1.4590	1.4330	1.3660	1.2830	1.2080
	0.05	1.2650	1.3500	1.4360	1.4740	1.4460	1.3750	1.2930	1.2220
	0.10	1.2640	1.3420	1.4380	1.4920	1.4680	1.3910	1.3050	1.2330
<i>r = 8</i>									
0.25	0.01	1.0100	0.9982	0.9978	0.9977	0.9978	0.9978	0.9978	0.9978
	0.05	1.0124	0.9991	0.9983	0.9981	0.9982	0.9983	0.9985	0.9988
	0.10	1.0144	1.0101	0.9989	0.9986	0.9987	0.9990	0.9994	0.9999
0.50	0.01	1.0382	1.0357	1.0358	1.0360	1.0360	1.0356	1.0350	1.0340
	0.05	1.0431	1.0383	1.0375	1.0377	1.0381	1.0382	1.0381	1.0379
	0.10	1.0481	1.0415	1.0399	1.0402	1.0409	1.0416	1.0421	1.0424
1.00	0.01	1.1240	1.1450	1.1600	1.1660	1.1630	1.1520	1.1360	1.1170
	0.05	1.1250	1.1450	1.1660	1.1760	1.1750	1.1630	1.1460	1.1280
	0.10	1.1320	1.1510	1.1750	1.1900	1.1900	1.1780	1.1590	1.1390
1.50	0.01	1.2180	1.2940	1.3530	1.3740	1.3550	1.3010	1.2330	1.1730
	0.05	1.2080	1.2800	1.3620	1.4030	1.3810	1.3190	1.2490	1.1890
	0.10	1.2130	1.2820	1.3790	1.4400	1.4190	1.3440	1.2650	1.2020
<i>r = 10</i>									
0.25	0.01	0.9990	0.9969	0.9966	0.9965	0.9965	0.9965	0.9966	0.9966
	0.05	1.0115	0.9980	0.9970	0.9968	0.9969	0.9971	0.9973	0.9976
	0.10	1.0138	0.9990	0.9976	0.9973	0.9974	0.9978	0.9983	0.9989
0.50	0.01	1.0330	1.0302	1.0301	1.0303	1.0303	1.0300	1.0295	1.0289
	0.05	1.0385	1.0332	1.0322	1.0324	1.0327	1.0330	1.0331	1.0332
	0.10	1.0441	1.0369	1.0350	1.0352	1.0360	1.0368	1.0376	1.0384
1.00	0.01	1.1020	1.1210	1.1350	1.1400	1.1380	1.1290	1.1150	1.0993
	0.05	1.1060	1.1240	1.1450	1.1560	1.1540	1.1440	1.1280	1.1120
	0.10	1.1160	1.1340	1.1590	1.1770	1.1770	1.1630	1.1450	1.1260
1.50	0.01	1.1740	1.2420	1.2970	1.3180	1.3020	1.2550	1.1990	1.1490
	0.05	1.1730	1.2380	1.3210	1.3660	1.3450	1.2850	1.2200	1.1660
	0.10	1.1850	1.2500	1.3560	1.4320	1.4060	1.3230	1.2430	1.1830

**Table 4.**  $RE(\hat{\theta}_4, Y_1)$  when  $\lambda = 0.5$ .

c	$\alpha$	$\delta$							
		0.4	0.6	0.8	1.0	1.2	1.4	1.6	1.8
<i>r = 6</i>									
0.25	0.01	1.0178	1.0157	1.0164	1.0178	1.0195	1.0215	1.0237	1.0263
	0.05	1.0225	1.0175	1.0172	1.0184	1.0204	1.0229	1.0260	1.0297
	0.10	1.0266	1.0193	1.0183	1.0193	1.0215	1.0246	1.0287	1.0338
0.50	0.01	1.1420	1.2100	1.6020	1.8800	1.2840	1.1820	1.1410	1.1220
	0.05	1.1650	1.2070	1.5070	2.0600	1.2900	1.1830	1.1430	1.1240
	0.10	1.1860	1.2130	1.4710	2.3000	1.3030	1.1890	1.1480	1.1310
1.00	0.01	2.3100	1.3070	1.1890	1.1470	1.1270	1.1180	1.1160	1.1230
	0.05	5.4400	1.3900	1.2090	1.1550	1.1330	1.1230	1.1220	1.1310
	0.10	28.100	1.4760	1.2320	1.1660	1.1400	1.1310	1.1320	1.1430
1.50	0.01	1.9600	1.3330	1.2220	1.1810	1.1630	1.1590	1.1720	1.2200
	0.05	2.9300	1.4100	1.2430	1.1900	1.1700	1.1670	1.1820	1.2350
	0.10	5.1900	1.4910	1.2680	1.2030	1.1800	1.1780	1.1980	1.2590
<i>r = 8</i>									
0.25	0.01	1.0163	1.0134	1.0138	1.0151	1.0168	1.0189	1.0214	1.0243
	0.05	1.0218	1.0152	1.0146	1.0157	1.0177	1.0204	1.0239	1.0285
	0.10	1.0266	1.0170	1.0155	1.0165	1.0188	1.0222	1.0269	1.0336
0.50	0.01	1.1410	1.2280	5.4400	1.2850	1.1500	1.1070	1.0872	1.0767
	0.05	1.1700	1.2200	2.3900	1.3110	1.1530	1.1090	1.0904	1.0812
	0.10	1.1950	1.2250	1.9890	1.3430	1.1600	1.1140	1.0956	1.0879
1.00	0.01	1.8350	1.2150	1.1340	1.1060	1.0918	1.0851	1.0834	1.0869
	0.05	3.1800	1.2740	1.1480	1.1110	1.0956	1.0895	1.0894	1.0954
	0.10	8.3200	1.3340	1.1640	1.1180	1.1010	1.0955	1.0974	1.1070
1.50	0.01	1.7190	1.2430	1.1620	1.1320	1.1190	1.1160	1.1230	1.1510
	0.05	2.4700	1.3030	1.1770	1.1390	1.1240	1.1220	1.1330	1.1670
	0.10	4.0100	1.3640	1.1950	1.1480	1.1320	1.1310	1.1460	1.1880
<i>r = 10</i>									
0.25	0.01	1.0156	1.0118	1.0120	1.0132	1.0149	1.0169	1.0195	1.0228
	0.05	1.0220	1.0136	1.0127	1.0137	1.0157	1.0185	1.0224	1.0278
	0.10	1.0276	1.0154	1.0136	1.0144	1.0167	1.0203	1.0258	1.0339
0.50	0.01	1.1410	1.2490	1.8260	1.1700	1.1020	1.0760	1.0633	1.0564
	0.05	1.1780	1.2340	2.9000	1.1820	1.1040	1.0785	1.0669	1.0616
	0.10	1.2090	1.2370	11.100	1.1970	1.1090	1.0825	1.0721	1.0687
1.00	0.01	1.6690	1.1670	1.1040	1.0822	1.0718	1.0666	1.0653	1.0678
	0.05	2.7200	1.2160	1.1150	1.0862	1.0748	1.0706	1.0712	1.0767
	0.10	5.4900	1.2650	1.1270	1.0917	1.0793	1.0761	1.0789	1.0881
1.50	0.01	1.6170	1.1930	1.1270	1.1040	1.0938	1.0911	1.0964	1.1160
	0.05	2.3300	1.2450	1.1400	1.1090	1.0982	1.0972	1.1060	1.1320
	0.10	3.8900	1.2970	1.1540	1.1160	1.1040	1.1050	1.1180	1.1540

**Table 5.**  $RE(\hat{\theta}_1, Y_1)$  when  $\lambda = 2.5$ .

c	$\alpha$	$\delta$							
		0.4	0.6	0.8	1.0	1.2	1.4	1.6	1.8
<i>r = 6</i>									
0.25	0.01	0.9980	0.9936	0.9928	0.9927	0.9927	0.9928	0.9930	0.9934
	0.05	1.0150	0.9954	0.9934	0.9931	0.9932	0.9936	0.9945	0.9958
	0.10	1.0235	0.9972	0.9939	0.9934	0.9937	0.9945	0.9961	0.9988
0.50	0.01	1.0209	1.0155	1.0142	1.0142	1.0143	1.0146	1.0153	1.0164
	0.05	1.0370	1.0244	1.0197	1.0194	1.0202	1.0221	1.0252	1.0291
	0.10	1.0568	1.0342	1.0247	1.0239	1.0260	1.0304	1.0372	1.0452
1.00	0.01	1.0478	1.0620	1.0838	1.0938	1.0901	1.0849	1.0801	1.0743
	0.05	1.0804	1.1060	1.2190	1.4390	1.3670	1.2500	1.1790	1.1350
	0.10	1.1200	1.1610	1.6850	1.7500	1.1300	2.2000	1.3500	1.2070
1.50	0.01	1.0735	1.1240	1.2470	1.3370	1.2840	1.2270	1.1870	1.1520
	0.05	1.1200	1.2210	3.1800	1.4120	1.5970	3.4200	1.5400	1.2600
	0.10	1.1750	1.3530	1.4610	1.2170	1.2690	1.6350	2.4000	1.3610
<i>r = 8</i>									
0.25	0.01	1.9997	0.9934	0.9923	0.9922	0.9922	0.9924	0.9927	0.9934
	0.05	1.0227	0.9958	0.9929	0.9925	0.9927	0.9933	0.9945	0.9970
	0.10	1.0405	0.9983	0.9933	0.9927	0.9931	0.9942	0.9967	0.0119
0.50	0.01	1.0229	1.0154	1.0131	1.0131	1.0132	1.0138	1.0151	0.0170
	0.05	1.0513	1.0289	1.0196	1.0189	1.0202	1.0238	1.0300	0.0381
	0.10	1.0898	1.0437	1.0248	1.0230	1.0263	1.0345	1.0490	0.0674
1.00	0.01	1.0491	1.0676	1.1230	1.1610	1.1470	1.1280	1.1070	0.0863
	0.05	1.1050	1.1560	2.3000	1.2370	1.2910	1.8060	1.5920	0.2120
	0.10	1.1760	1.2790	1.3180	1.1360	1.1550	1.2870	3.3900	0.3670
1.50	0.01	1.0739	1.1440	3.3800	1.8200	3.1300	2.2000	1.3700	0.1900
	0.05	1.1530	1.3900	1.2040	1.1260	1.1480	1.2460	2.2400	0.4300
	0.10	1.2480	1.8240	1.1670	1.1030	1.1210	1.2090	1.8510	0.6260
<i>r = 10</i>									
0.25	0.01	1.0129	0.9935	0.9920	0.9919	0.9919	0.9921	0.9927	0.9937
	0.05	1.0376	0.9967	0.9925	0.9921	0.9923	0.9931	0.9950	0.9993
	0.10	1.0752	1.0100	0.9929	0.9923	0.9927	0.9940	0.9977	1.0175
0.50	0.01	1.0284	1.0172	1.0131	1.0129	1.0132	1.0142	1.0166	1.0201
	0.05	1.0797	1.0369	1.0199	1.0184	1.0204	1.0265	1.0385	1.0559
	0.10	1.1540	1.0586	1.0246	1.0216	1.0260	1.0392	1.0675	1.1090
1.00	0.01	1.0584	1.0881	1.5470	2.2200	3.0100	1.6490	1.2300	1.1220
	0.05	1.1540	1.2800	1.1670	1.0926	1.1040	1.1640	1.6570	1.4010
	0.10	1.2770	1.5740	1.1450	1.0784	1.0894	1.1500	1.5380	1.7050
1.50	0.01	1.0865	1.2150	1.2080	1.1400	1.1620	1.2400	2.1100	1.3070
	0.05	1.2180	2.2000	1.1090	1.0740	1.0851	1.1300	1.3540	1.8710
	0.10	1.3700	6.3400	1.1090	1.0692	1.0822	1.1410	1.4590	2.0800

**Table 6.**  $RE(\hat{\theta}_2, Y_1)$  when  $\lambda = 2.5$ .

c	$\alpha$	$\delta$							
		0.4	0.6	0.8	1.0	1.2	1.4	1.6	1.8
<i>r = 6</i>									
0.25	0.01	0.9980	0.9936	0.9928	0.9927	0.9927	0.9928	0.9930	0.9934
	0.05	1.0150	0.9954	0.9934	0.9931	0.9932	0.9936	0.9944	0.9957
	0.10	1.0235	0.9972	0.9939	0.9934	0.9937	0.9945	0.9960	0.9986
0.50	0.01	1.0209	1.0155	1.0142	1.0141	1.0142	1.0144	1.0145	1.0148
	0.05	1.0370	1.0244	1.0197	1.0193	1.0201	1.0216	1.0236	1.0261
	0.10	1.0568	1.0342	1.0248	1.0238	1.0258	1.0296	1.0347	1.0407
1.00	0.01	1.0478	1.0622	1.0831	1.0907	1.0878	1.0760	1.0606	1.0475
	0.05	1.0804	1.1070	1.2210	1.3860	1.3280	1.1920	1.1170	1.0836
	0.10	1.1200	1.1630	1.7200	1.9140	2.7300	1.5400	1.1990	1.1280
1.50	0.01	1.0736	1.1250	1.2400	1.2970	1.2600	1.1700	1.1010	1.0645
	0.05	1.1200	1.2250	3.0700	1.4780	1.7590	1.7780	1.1920	1.1080
	0.10	1.1750	1.3610	1.4470	1.2300	1.2960	3.1300	1.3130	1.1590
<i>r = 8</i>									
0.25	0.01	1.9997	0.9934	0.9923	0.9922	0.9922	1.9924	0.9927	0.9933
	0.05	1.0227	0.9958	0.9929	0.9925	0.9927	1.9933	0.9945	0.9968
	0.10	1.0405	0.9983	0.9933	0.9927	0.9931	1.9942	0.9966	0.0116
0.50	0.01	1.0229	1.0155	1.0131	1.0130	1.0132	1.0135	1.0141	1.0152
	0.05	1.0513	1.0289	1.0196	1.0188	1.0201	1.0230	1.0276	1.0336
	0.10	1.0898	1.0438	1.0249	1.0229	1.0261	1.0332	1.0447	1.0595
1.00	0.01	1.0491	1.0678	1.1220	1.1520	1.1400	1.1020	1.0692	1.0501
	0.05	1.1050	1.1570	2.2000	1.2560	1.3250	3.9100	1.2000	1.1130
	0.10	1.1760	1.2820	1.3090	1.1400	1.1640	1.4480	1.4680	1.1970
1.50	0.01	1.0739	1.1450	3.0300	2.1900	8.6200	1.3430	1.1180	1.0661
	0.05	1.1530	1.3990	1.2020	1.1310	1.1570	1.4250	1.3410	1.1380
	0.10	1.2480	1.8530	1.1650	1.1050	1.1270	1.2930	1.7330	1.2260
<i>r = 10</i>									
0.25	0.01	1.0129	1.9935	1.9920	1.9919	1.9919	1.9921	1.9926	1.9936
	0.05	1.0376	1.9967	1.9925	1.9921	1.9923	1.9931	1.9949	1.9991
	0.10	1.0752	1.0100	1.9929	1.9923	1.9927	1.9940	1.9976	1.0171
0.50	0.01	1.0284	1.0172	1.0131	1.0129	1.0131	1.0138	1.0153	1.0176
	0.05	1.0797	1.0370	1.0199	1.0183	1.0203	1.0253	1.0346	1.0483
	0.10	1.1540	1.0587	1.0246	1.0215	1.0257	1.0372	1.0601	1.0947
1.00	0.01	1.0584	1.0884	1.5360	3.1200	3.9100	1.2490	1.0981	1.0610
	0.05	1.1540	1.2830	1.1650	1.0952	1.1090	1.2380	1.5580	1.1730
	0.10	1.2770	1.5820	1.1430	1.0799	1.0925	1.1910	5.7400	1.3220
1.50	0.01	1.0866	1.2170	1.2090	1.1480	1.1740	1.6640	1.1760	1.0775
	0.05	1.2180	2.2500	1.1080	1.0758	1.0884	1.1770	2.0700	1.1970
	0.10	1.3700	7.2000	1.1080	1.0703	1.0850	1.1810	6.4600	1.3390

**Table 7.**  $RE(\hat{\theta}_3, Y_1)$  when  $\lambda = 2.5$ .

c	$\alpha$	$\delta$							
		0.4	0.6	0.8	1.0	1.2	1.4	1.6	1.8
<i>r = 6</i>									
0.25	0.01	0.9980	0.9936	0.9928	0.9927	0.9927	0.9928	0.9930	0.9934
	0.05	1.0150	0.9954	0.9934	0.9931	0.9932	0.9936	0.9944	0.9957
	0.10	1.0235	0.9972	0.9939	0.9934	0.9937	0.9945	0.9960	0.9987
0.50	0.01	1.0208	1.0154	1.0142	1.0142	1.0143	1.0145	1.0148	1.0155
	0.05	1.0370	1.0243	1.0197	1.0194	1.0202	1.0218	1.0243	1.0273
	0.10	1.0567	1.0340	1.0247	1.0239	1.0260	1.0300	1.0357	1.0425
1.00	0.01	1.0475	1.0609	1.0836	1.0929	1.0896	1.0792	1.0669	1.0562
	0.05	1.0801	1.1040	1.2190	1.4210	1.3610	1.2160	1.1370	1.1010
	0.10	1.1200	1.1580	1.6840	1.7970	2.1700	1.7560	1.2450	1.1540
1.50	0.01	1.0730	1.1200	1.2440	1.3260	1.2790	1.1890	1.1240	1.0865
	0.05	1.1200	1.2140	3.2200	1.4290	1.6140	2.5700	1.2670	1.1470
	0.10	1.1750	1.3420	1.4610	1.2210	1.2710	1.9470	1.4780	1.2120
<i>r = 8</i>									
0.25	0.01	0.9997	0.9934	0.9923	0.9922	0.9922	0.9924	0.9927	0.9933
	0.05	1.0226	0.9958	0.9929	0.9925	0.9927	0.9933	0.9945	0.9969
	0.10	1.0405	0.9983	0.9933	0.9927	0.9931	0.9942	0.9966	1.0117
0.50	0.01	1.0229	1.0154	1.0131	1.0131	1.0132	1.0136	1.0145	1.0160
	0.05	1.0513	1.0287	1.0196	1.0188	1.0202	1.0233	1.0285	1.0355
	0.10	1.0897	1.0435	1.0248	1.0230	1.0263	1.0338	1.0464	1.0627
1.00	0.01	1.0489	1.0659	1.1220	1.1580	1.1450	1.1110	1.0810	1.0619
	0.05	1.1050	1.1510	2.3500	1.2430	1.2970	3.7600	1.2800	1.1430
	0.10	1.1760	1.2700	1.3200	1.1370	1.1570	1.3550	1.8480	1.2480
1.50	0.01	1.0736	1.1370	3.0600	1.8940	3.7500	1.4730	1.1660	1.0953
	0.05	1.1520	1.3650	1.2050	1.1280	1.1500	1.3200	1.7700	1.2020
	0.10	1.2470	1.7580	1.1680	1.1040	1.1220	1.2460	1.9000	1.3190
<i>r = 10</i>									
0.25	0.01	1.0129	0.9935	0.9920	0.9919	0.9919	0.9921	0.9926	0.9937
	0.05	1.0375	0.9967	0.9925	0.9921	0.9923	0.9931	0.9949	0.9992
	0.10	1.0752	1.0100	0.9929	0.9923	0.9927	0.9940	0.9977	0.0173
0.50	0.01	1.0284	1.0171	1.0131	1.0129	1.0131	1.0140	1.0158	1.0187
	0.05	1.0797	1.0367	1.0198	1.0184	1.0204	1.0258	1.0362	1.0515
	0.10	1.1540	1.0582	1.0245	1.0216	1.0259	1.0381	1.0631	1.1010
1.00	0.01	1.0583	1.0852	1.5180	2.3800	8.9500	1.3270	1.1300	1.0799
	0.05	1.1540	1.2660	1.1690	1.0934	1.1060	1.2000	3.9100	1.2340
	0.10	1.2770	1.5400	1.1460	1.0789	1.0900	1.1700	2.4700	1.4270
1.50	0.01	1.0863	1.1990	1.2120	1.1420	1.1670	1.3930	1.3610	1.1220
	0.05	1.2180	1.9920	1.1090	1.0745	1.0860	1.1530	2.3600	1.3110
	0.10	1.3690	4.4800	1.1100	1.0695	1.0828	1.1600	2.1800	1.4960

**Table 8.**  $RE(\hat{\theta}_4, Y_1)$  when  $\lambda = 2.5$ .

c	$\alpha$	$\delta$							
		0.4	0.6	0.8	1.0	1.2	1.4	1.6	1.8
<i>r = 6</i>									
0.25	0.01	1.0267	1.0065	1.0053	1.0058	1.0069	1.0086	1.0117	1.0177
	0.05	1.0578	1.0092	1.0058	1.0060	1.0075	1.0105	1.0171	1.0330
	0.10	1.0904	1.0118	1.0064	1.0063	1.0082	1.0127	1.0234	1.0533
0.50	0.01	1.2480	1.6250	1.0662	1.0335	1.0243	1.0202	1.0187	1.0192
	0.05	1.4060	1.3910	1.0831	1.0351	1.0255	1.0229	1.0242	1.0294
	0.10	1.5160	1.3740	1.1010	1.0372	1.0272	1.0261	1.0305	1.0421
1.00	0.01	2.3400	1.0675	1.0327	1.0255	1.0227	1.0218	1.0229	1.0265
	0.05	2.6100	1.1060	1.0375	1.0266	1.0242	1.0254	1.0305	1.0424
	0.10	2.7500	1.1450	1.0424	1.0282	1.0261	1.0295	1.0393	1.0621
1.50	0.01	2.4200	1.0817	1.0416	1.0332	1.0303	1.0302	1.0338	1.0441
	0.05	2.7100	1.1260	1.0474	1.0347	1.0325	1.0355	1.0456	1.0716
	0.10	2.7900	1.1720	1.0534	1.0367	1.0352	1.0414	1.0593	1.1070
<i>r = 8</i>									
0.25	0.01	1.0478	1.0060	1.0042	1.0045	1.0054	1.0071	1.0106	1.0188
	0.05	1.1200	1.0094	1.0047	1.0047	1.0060	1.0092	1.0174	1.0428
	0.10	1.1960	1.0126	1.0052	1.0050	1.0067	1.0115	1.0258	1.0776
0.50	0.01	1.3780	1.9530	1.0462	1.0239	1.0177	1.0152	1.0150	1.0171
	0.05	1.6060	1.4730	1.0586	1.0250	1.0188	1.0181	1.0216	1.0317
	0.10	1.7270	1.4480	1.0711	1.0265	1.0203	1.0215	1.0294	1.0510
1.00	0.01	2.9300	1.0607	1.0247	1.0190	1.0170	1.0168	1.0187	1.0243
	0.05	2.5000	1.1050	1.0289	1.0198	1.0184	1.0206	1.0279	1.0467
	0.10	2.2200	1.1520	1.0332	1.0209	1.0201	1.0248	1.0388	1.0773
1.50	0.01	2.9100	1.0737	1.0315	1.0248	1.0227	1.0233	1.0277	1.0403
	0.05	2.5000	1.1260	1.0367	1.0259	1.0247	1.0288	1.0418	1.0793
	0.10	2.2200	1.1830	1.0419	1.0273	1.0271	1.0349	1.0587	1.1350
<i>r = 10</i>									
0.25	0.01	1.0934	1.0060	1.0034	1.0037	1.0045	1.0062	1.0102	1.0216
	0.05	1.2490	1.0102	1.0040	1.0039	1.0051	1.0085	1.0189	1.0596
	0.10	1.3940	1.0145	1.0045	1.0041	1.0057	1.0110	1.0300	1.1180
0.50	0.01	1.5560	2.3000	1.0359	1.0186	1.0139	1.0124	1.0132	1.0173
	0.05	1.7900	1.5490	1.0462	1.0194	1.0150	1.0156	1.0214	1.0387
	0.10	1.8780	1.5220	1.0566	1.0205	1.0164	1.0192	1.0312	1.0698
1.00	0.01	2.9000	1.0596	1.0200	1.0151	1.0136	1.0139	1.0168	1.0249
	0.05	2.1600	1.1130	1.0240	1.0157	1.0150	1.0180	1.0281	1.0580
	0.10	2.0700	1.1750	1.0279	1.0166	1.0165	1.0225	1.0419	1.1080
1.50	0.01	2.8500	1.0726	1.0256	1.0198	1.0183	1.0194	1.0249	1.0415
	0.05	2.1600	1.1370	1.0304	1.0206	1.0201	1.0252	1.0422	1.0997
	0.10	2.0700	1.2130	1.0353	1.0218	1.0223	1.0317	1.0637	1.1940

than the underestimation.

### 3.1. When $\lambda = 0.5$

From these tables it is observed that the shrinkage estimators  $\hat{\theta}_4$  perform better than the improved estimator  $Y_1$  for all considered values of  $r, c, \delta, \lambda$  and  $\alpha$ . The estimators  $\hat{\theta}_1$ ,  $\hat{\theta}_2$  and  $\hat{\theta}_3$  perform better than  $Y_1$  when  $c \geq 0.5$ . The estimators  $\hat{\theta}_4$  attain maximum efficiency at the point  $\delta = 0.4$  and others near to the point  $\delta = 1$ .

For fixed  $c$  and level of significance  $\alpha$ , as the uncensored sample size  $r$  increases, the relative efficiency decreases in all considered values of  $\delta$  for all the estimators.

For fixed  $r$  and  $\alpha$ , when  $c$  increases the relative efficiency increases in all considered values of  $\delta$  for all estimators.

It has been seen that as the level of significance  $\alpha$  increases the relative efficiency increases in all considered values of  $\delta$  for all estimators.

### 3.2. When $\lambda = 2.5$

From these tables it is observed that the shrinkage estimators  $\hat{\theta}_4$  perform better than the improved estimator  $Y_1$  for all considered values of  $r, c, \delta, \lambda$  and  $\alpha$ . The estimators  $\hat{\theta}_1$ ,  $\hat{\theta}_2$  and  $\hat{\theta}_3$  perform better than  $Y_1$  when  $c \geq 0.5$ . The estimators  $\hat{\theta}_4$  attain maximum efficiency at the point  $\delta = 0.4$  and others near to the point  $\delta = 1.6$ .

For fixed  $c$  and level of significance  $\alpha$ , as the uncensored sample size  $r$  increases, the relative efficiency decreases in the region  $0.8 \leq \delta \leq 1.6$  for the estimator  $\hat{\theta}_1$ , and in the region  $0.8 \leq \delta \leq 1.4$  for the estimators  $\hat{\theta}_2$  and  $\hat{\theta}_3$ , and also for estimator  $\hat{\theta}_4$  it decreases for all considered values of  $\delta$ .

For fixed  $r$  and  $\alpha$ , when  $c$  increases the relative efficiency increases in all considered values of  $\delta$  for all estimators.

It has been seen that as the level of significance  $\alpha$  increases the relative efficiency increases in  $1.6 \leq \delta \leq 0.6$  and also in  $0.8 \leq \delta \leq 1.4$  when  $c = 0.25, 0.5$ , and decreases for  $0.8 \leq \delta \leq 1.4$  when  $c \geq 1.5$  and  $r \geq 8$  for estimators  $\hat{\theta}_1$ ,  $\hat{\theta}_2$  and  $\hat{\theta}_3$ , and for estimator  $\hat{\theta}_4$  it increases for all considered values of  $\delta$ .

## 4. Recommendations

In this study, some shrinkage estimators ( $\hat{\theta}_1$ ,  $\hat{\theta}_2$ ,  $\hat{\theta}_3$  and  $\hat{\theta}_4$ ) for the scale parameter of a gamma distribution, when type-II censored data are available, suggested that under the ILL loss function assuming the shape parameter was to be known. The comparisons of the proposed estimators made with improved estimator  $Y_1$ . The recommendations have been presented, based on the relative

efficiency for all the shrinkage estimator. From the previous observations, the shrinkage estimators  $\hat{\theta}_4$  perform better than the improved estimator  $Y_1$  for all considered values of  $r, c, \delta, \lambda$  and  $\alpha$ , and the estimators  $\hat{\theta}_1$ ,  $\hat{\theta}_2$  and  $\hat{\theta}_3$  perform better than  $Y_1$  when  $c \geq 0.5$ . Since the shrinkage estimators  $\hat{\theta}_4$  always perform better than other shrinkage estimators if the gain in efficiency does not matter, therefore we strongly suggest using the shrinkage estimators  $\hat{\theta}_4$  for the scale parameter of a gamma distribution, when Type-II censored data are available, suggested under the ILL loss function.

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