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This paper provides a few initial results from a simulation study of raking ratio estimators. The particular aspects of raking examined here were in part suggested by our various "real life" experiences with raking, some of which are described in the previous paper [31].

1. SOME UNRESOLVED ISSUES

Despite the extensive literature on raking (see references in [31]), there remain a number of major unresolved issues. With one slight exception [35], for example, there has been no consideration of what happens to raking ratio estimators when the outside (or external) marginal totals are themselves subject to error. There also seems to have been virtually no study made of the properties of raking estimators when the sample (or interior) marginal totals are subject to misclassification or other response problems.

Perhaps the issue of most concern to us has been the performance of raking ratio estimators when used, as is commonly the case, in surveys (like the Current Population Survey [49])which suffer from coverage errors. In particular, what kind of bias-variance tradeoffs can be expected (especially in small samples)?

The use of raking to make coverage adjustments is the focus of the simulation results provided here. Our attention will be confined to an examination of raking's impact on the mean square error when adjusting for coverage errors in samples of small to moderate size. We will assume basically that the coverage errors encountered are such that every class of individuals in the population is represented in the sample, but not necessarily in its proper proportion.

2. INITIAL SIMULATIONS

The simulation experiments we have conducted so far systematically vary six different factors. These factors, and how each was treated, are discussed briefly below, along with some of the hypotheses we wished to test.

<u>Sample size</u>.--Four different sample sizes were examined: n=50; n=100; n=400; and n=800. One of the hypotheses of interest here was that for the smaller samples (e.g., n=50 or n=100) there would actually be a marked deterioration in variance performance over the unraked estimator. (This turns out to be the case in at least some circumstances as we will see below.)

Number of levels of each marginal raked.--Attention was confined in the simulation to only the simplest form of univariate raking, i.e., the case where we successively rake a sample to known outside row and column totals. In the computations done for these <u>Proceedings</u>, we have looked at just 4x4 and 7x7 tables. 1/ We hypothesized that the variance performance of the 7x7 raked data would be inferior to that for the 4x4 case in small samples. We surmised, too, that it would become superior only in moderate to large samples where there was a very strong dependence between the raked and unraked information.

Whether the expected sample totals and outside marginal totals were equal. -- Two alternatives were considered in the simulation (see figure 1):

- (a) <u>Unbiased</u>.--The row and column outside marginal totals, or "controls," were equal to their corresponding expected sample counts.
- (b) <u>Biased</u>.--The controls were taken to be different from the sample expected values.

The questions of obvious interest here are to what extent did the bias "correction" adversely affect variances, and at what point did it begin to reduce the mean square error?

Figure	1Expected	Sample M	Marginals	and Outside
	Controls	for Samp	ple Size ı	n ≈ 50

Row or	Unbiased			Biased		
Column Class	7 x 7	4 x 4	_	7 x 7	4 x 4	
First	10.0		(5.0)	
Second	10.0	20.0	١.	5.0	10.0	
Third	7.5	1	1	5.0	1	
Fourth	7.5	15.0	1	7.5) 12.5	
Fifth	5.0	10.0	5	7.5	17.5	
Sixth	5.0	٠	ł	10.0) 17.5	
Seventh	5.0	5.0		10.0	10.0	

Note: For the larger sample sizes considered in the simulations the "controls" used were multiples (2,8 and 16) of those shown above (for n = 100, 400 and 800 respectively).

Extent of relationships within variables being raked. 2/--Three alternatives were considered (see figure 2):

- (a) <u>Totally unrelated</u>.--The row and column variables were statistically independent.
- (b) <u>Totally related</u>.--The row and column variables were the same (i.e., we set the column variable equal to that for the row). This is equivalent to employing a <u>simple</u> <u>ratio</u> estimator based on the rows.
- (c) <u>Partially related</u>.--At random, one-fourth of the time, we made the column variable equal to that for the row.

We hypothesized that when the raked variables were totally unrelated, the adjustment would have a greater impact on the variance than for the partially or totally related settings.

Figure 2 - Alternative relationships between row and column variables used in the raking

Extent of Relationship	Row Variable X ₁	Column Variable X ₂
Totally Unrelated	$\widetilde{x}_1 = x_1$	x ₂ = x ₂
Totally Related	$\widetilde{x}_1 = x_1$	$\widetilde{\mathbf{x}}_2 = \mathbf{x}_1$
Partially Related	$\widetilde{\mathbf{x}}_1 = \mathbf{x}_1$	$\widetilde{\mathbf{x}}_2 = \begin{cases} \mathbf{x}_1 & \mathbf{x}_4 < .25 \\ \\ \mathbf{x}_2 & \mathbf{x}_2 & 2.25 \end{cases}$

Nature of variables whose mean square error we wish to reduce.--The basic structure of the experiment was to draw samples of vectors,

$$\underline{\mathbf{X}}' = (\mathbf{X}_1, \mathbf{X}_2, \mathbf{X}_3, \mathbf{X}_4),$$

each component of which was an independent uniform random number on (0, 1). Two basic functions,

$$Y_h = g_h(\underline{X})$$
 h = 1, 2,

were examined in the simulations. These were:

- (a) <u>bounded</u> "uniform type" random variables $\{g_1(\underline{X})\}\$ constructed essentially as linear combinations of some of the components of \underline{X} ; and
- (b) <u>unbounded</u> "exponential type" random variables $\{g_2(\underline{X})\}$ obtained as linear combina-

tions of some of the components of \underline{Z} where \underline{Z} and \underline{X} are related component for component by the (probability integral) transformation

$$X = 1 - \exp \left\{-(0.1) \text{ times } Z\right\}.$$

Naturally, greater variance effects, both increases and decreases, were anticipated for the unbounded "exponential type" variables rather than for the bounded "uniform" ones.

Degree of dependence between raked and unraked variables.--There were three types of dependence considered between the raked row and column variables and the "unraked" { Y_h } variables. We have characterized these as "Complete Independence," "Complete Dependence," and "Partial Dependence." There are two versions of each form of dependence subject to whether a "uniform" or "exponential" type variable is being looked at. (See figure 3.)

<u>Replication of experimental conditions.--The</u> results to be discussed in the next section were based on 200 replications for the n=50 case and 100 replications for sample sizes n=100, n=400 and n=800. In every instance the raking was carried out until for all levels of the row and column marginal either

or the process had proceeded for 100 cycles.

Generally, except for samples of size n=50, convergence occurred quickly even when the expected sample marginals differed from the outside totals being introduced. For the n=50 samples, two difficulties arose. First, in five or six cases, one (or more) row or column classes of the data were zero and, hence, raking could not be carried out unless some collapsing was done. (We discarded these samples before raking and they were not used in any of the comparisons.) Second, again because n=50 is so small, for the biased 7x7case a substantial portion of the replications were ones where we proceeded the full 100 cycles, i.e., convergence did not occur. (These latter samples were, however, still used in the comparisons, and undoubtedly contributed to the poor variance performance of the raking estimator for n=50.)

Figure	3Altern	ative	e relat:	ionships	between	the
	raked	and u	ınraked	variable	25	

Form of Dependence	Uniform-type variables	Exponential-type variables
Complete Independence	$Y = X_3$	¥ = Z ₃
Complete Dependence	$Y = 0.5 + \sqrt{9/5} \{1/3(X_15) + 2/3(X_25)\}$	$Y = 10 + \sqrt{9/5} \{ 1/3(Z_1 - 10) + 2/3(Z_2 - 10) \}$
Partial Dependence	$Y=0.5+\sqrt{36/14} \left\{ \frac{1}{3}(x_15) +\frac{1}{6}(x_25) +\frac{1}{2}(x_35) \right\}$	$Y=10+\sqrt{36/14} \frac{1}{3} \frac{1}{3} \frac{1}{2} \frac{1}{3} \frac{1}{3$

3. RESULTS OF INITIAL SIMULATIONS

In this section we will attempt to highlight, factor by factor, the simulation results obtained. We will focus our remarks (see tables 1 to 3) solely on the performance characteristics of the means of the unconstrained variables { Y_h }.

<u>Sample size</u>.--For the n=50 case, as expected, there was an increase in the variance caused by the raking. This was especially marked if an adjustment for coverage errors was being made. It is also interesting to note that the increase occurred quite generally, even sometimes when the raked and unraked variables were completely dependent.

For the larger sample sizes, some variance "price" continued to be paid when raking if the unconstrained mean was independent of the raked variables. When it was not independent, substantial benefits in reduced variance were achievable.

Number of marginal totals.--As hypothesized, for n=50 the variance performance of the 7x7 raked estimator was inferior to the 4x4 one. If the raked and unraked variables were independent, the 7x7 raked estimator continued to be inferior for larger samples. The difference decreased as "n" grew larger but did not disappear even for n=800. If the raked and unraked variables were dependent, then the performance of the 7x7 and 4x4 estimators followed no consistent pattern.

To compensate for the variance increase that sometimes accompanied the use of more controls, there was an accompanying increase in the 7x7estimator's ability to reduce the coverage bias. Except for n=50, in fact, when adjusting for coverage errors, the root mean square error of the 7x7 estimator was smaller than the corresponding 4x4 estimator. This was true even though the classifiers used were such that the ratios of outside control to expected sample total were the same before and after collapsing from seven classes to four (see figure 1).

Whether expected sample totals and outside "controls" were equal. -- The variance impact of using raking as a coverage adjustment procedure can be clearly seen when we contrast its behavior to a raking estimator for which the outside totals and expected sample marginals were equal. Our simulation results show, for instance, that there was virtually always some increase in the variance when the controls differed from the expected sample size. This increase tended to be quite large when the unconstrained mean was independent of the raked variables. It diminished in importance in the partial dependence case and all but disappeared (for n>50) when there was complete dependence of the unconstrained mean on the raked variables.

Extent of relationship within variables being raked.--By and large, if the raked variables were not related to each other, then the raking had a greater impact on the variance. This impact could be either beneficial or adverse. If the unconstrained mean was independent of the raked variables, then the impact was adverse. On the other hand, if the unconstrained mean depended completely on the raked variables, then the greatest variance reductions were achieved.

Nature of variables being studied.--The same overall patterns we have been describing occurred for both the "uniform" and "exponential" type unconstrained means. There was some difference in the behavior of these two types of variables but it was a question of degree only. Substantially larger changes occurred in the "exponential" means than in the "uniform" ones when adjusting for bias. Conversely (contrary to our expectations), the variance impact of the raking tended to be smaller for "exponential" variables than for "uniform" ones (i.e., smaller relative increases or decreases occurred for the "exponential" cases, all other things being equal).

Degree of dependence between raked and unraked variables.--Perhaps the most important factor in deciding whether to use a raking ratio estimator is the degree of dependence anticipated between the raked and unraked variables. If there is little or no dependence, then raking just tends to increase the variance, especially in a very small samples or when attempting to correct for coverage errors. If there is a moderate amount of dependence, then raking can be quite beneficial. In fact, it can simultaneously reduce both the coverage bias and the sampling variance.

> 4. CONCLUSIONS AND AREAS FOR FUTURE STUDY

In a multi-purpose survey with many, many variables, we typically would have a situation in which a raking coverage adjustment materially reduced the mean square error of some variables. At the same time, however, the variance of the remainder would increase, possibly quite substantially, if the coverage errors were at all serious. It is hoped that the simulation results just described can aid practitioners in assessing the trade-offs involved in such settings. Obviously, though, this paper is just the beginning of our attempts to understand more about the performance of raking estimators when used to make coverage adjustments. For example, we need to go on and explore the behavior of alternative variance estimation procedures (as suggested by the discussant when we delivered this paper). We also want to see how much difference there is in the mean square error if we iterate only a few cycles instead of attempting to achieve complete convergence.

ACKNOWLEDGEMENTS

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FOOTNOTES

- 1/ When this paper was presented in San Diego, we also provided results for 5x5 and 6x6 tables. The patterns exhibited by all the simulations were roughly the same; hence, we have restricted ourselves just to the 4x4 and 7x7 tables, i.e., to the extremes.
- 2/ In our original paper we did not include this factor. The discussant suggested that at a minimum we compare the raked and simple ratio estimators as well as raked versus unraked estimators. We were happy to oblige.

		DEGREW OF DEPENDENCE BETWEEN RAKED VARIABLES AND UNCONSTRAINED MEAN								
ITEM	COME	PLETE INDEPEND	DENCE	COM	IPLETE DEPENDE	INCE	PAI	KTIAL DEPENDEN	iC ti	
215.9	RAKED MI	ARG INALS			RAKED MARGINALS		RAKED MARGINALS		SIMPLE	
	UNRELATED	RELATED	RATIO ADJUSTMENT	UNRELATED	RELATED	ADJUSTMENT	JNRELATED	RELATED	RATIO ADJUSIMEN	
UNBIASED CASE				PART I - 1	UN.I.FORM-TYPE	VARIABLES				
×7 RAKED DAT4 BY										
SAMPLE STZE 50 100 400 800	0.8542 0.1211 0.1203 0.0014	1.0635 0.2206 0.1235 0.0235	0.2240 0.0289 0.1438 0.0253	2.3136 0.1410 0.1758 0.1794	2.0679 0.4796 0.6536 0.6550	0.7365 0.4330 0.3799 0.4298	0,5305 0,5066 0,5139 0,4208	0.1573 0.6921 0.6611 0.5443	1.1651 0.5130 0.5958 0.5163	
×4 RAKED DATA BY SAMPLE SIZE 50	0.2748	0.4339	_0.0080	- 0.0945	0.0260	0.5403	0.0800	u.47 36	0.7513	
1.00 400 800	-0.0468 0.0673 -0.0211	0.1888 0.0806 0.0079	0.0297 0.0643 0.0135	0.2760 0.3408 0.3340	0.5813 0.7570 0.7714	0.1742 0.2856 0.3424	0.3324 0.4498 0.3730	J.6160 0.5835 0.5163	0.3150 0.4244 0.4037	
BIASED CASE										
227 RAKED DATA BY SAMPLE ST29 50 100 400 900	0.3377 0.0968 -0.3694 -0.6183	0.7194 0.3035 0.1404 0.4002	0.1969 70.0325 0.0183 70.1175	33.6511 41.4355 41.5663 41.5615	24.4543 28.8568 29.5871 29.7025	13.9715 13.8952 13.9881 14.0438	21.1873 25.3351 25.0019 24.7989	19.9535 21.7162 21.5794 21.4066	17.3114 16.9212 17.0000 16.9085	
×4 RAKED DATA BY	0. 100	001001			2	1.0100	2101100	21.1000	100000	
4 04450 044 51 SAMPLE ST28 50 100 400 809	0.0285 0.2712 0.4144 0.5504	0.1362 0.0143 0.1836 0.3188	T0.1143 T0.0826 0.0404 T0.0018	27.6979 39.7580 39.9650 39.9702	27.1578 28.1971 28.6269 28.7340	13.2422 13.0916 13.2998 13.3653	22,8496 23,9825 23,9736 23,8665	20.4130 20.7308 20.7322 20.6577	16.2613 16.0911 16.2547 16.2436	
	PART II - "EXPORENTIAL-TYPE" FARIABLES									
UNBIASED CASE										
×7 RAKED DATA BY SAMPLE SIZE 50 100 400	1.3634 0.1989 0.1886	1.4368 0.1502 0.2208	0.4899 0.0586 0.1949	-2.7747 0.5045 0.4704	-2.6520 0.7967 1.0457	1.5400 0.2276 0.2623	0,6020 0,5165 0,5880	-0.2711 0.5434 0.7766	2.1269 0.2449 0.5126	
800 ×4 <i>RAKSD DATA BY</i>	0.0165	0.0281	0.0394	0.4884	1.0333	0.3851	0,4448	0.5609	Ú. 4271	
SAMPLE SIZE 50 100 400	0.3756 0.0017 0.1093	0.4388 0.2128 0.1595	0.1071 -0.0780 0.0219	0,0216 0,6431 0,6575	0.4005 0.8924 1.1933	1.3722 0.0588 0.1901	0.0115 0.3490 0.5201	0.2325 0.5738 0.7190	1.7486 0.1018 0.3479	
800	0.0069	0.0262	0.0416	0.6844	1.2164	0.3189	0.4417	0.6174	0,3598	
BIASED CASS ×7 RAKED DATA BY										
SAMPLE SIZE 50 100 400. 200	0.1573 0.7540 0.0001 0.4199	0.9231 0.8549 0.4224 0.0895	0.7479 0.0790 0.2679 0.0870	56,9549 69,8573 70,2186 70,2537	41.2274 47.9980 49.4622 49.6454	24.3113 23.0271 23.4715 23.6101	35.1305 42.4058 42.0293 41.6778	32.7998 35.8049 36.1222 35.7550	29.4224 27.7306 28.1199 28.0209	
×4 RAKED DATA BY SAMPLE SIZE										
50. 100. 400. 800.	0.1166 0.2176 0.1266 0.2940	0.2331 0.3769 0.2678 0.0157	0.1852 0.0732 0.2689 0.2488	62,2009 66,7069 67,2370 67,3775	44.2236 46.8666 47.7921 48.0134	23.2652 21.7250 22.2779 22.4516	37,1619 40,0787 40,1911 40,1132	33.0016 34.1922 34.6547 34.5158	28.0104 26.4520 26.8656 26.9112	

-		DEGREE OF DEPENDENCE BETVESW RAKED VARIABLES AND UNCONSTRAINED MEAN								
T TEM	COM	PLETE INDEPEN	DENCE	CO	COMPLETS DEPENDENCE			PARTIAL DEPENDENCE		
	RAKED MARGIVALS		SIMPLE RATIO	RAKED MARGINALS		SIMPLE RATIO	RAKED MARGINALS		SIMPLS RATIO	
	UNRELATED	RELATED	ADJUSTMENT	UNRELATED	RELATED	ADJUSTNENT	JNRELATED	KELATSD	ADJ JST 18N	
				PART I -	'UNIFORM-TYPE	VARIABLES				
UNBIASED CASE										
×7 RAKRO DATA BY SAMPLE SIZE										
50 190	28.0885 9.7653	23.9756 10.6406	17.9181 7.5243	-3,9858 -83,0342	1.2444 140.8634	7.5032 14.2655	14.0071 19.4723	8.7258 16.7612	7.9716 17.1507	
400 200	_5.1158 _0.0951	4.4126 1.4903	1.4203 0.2960	_84.0396 _85.3595	_50,7409 _50,4629	14,5894 15,3240	19.7920 28.7347	20.9620 24.2963	22.1189 23.0474	
×4 RAKED DATA BY SAMPLE SIZE										
50 100	12.83F8 5.0E09	12.2242 F.1514	4.5984 3.2822	_3.º215 _69.5525	0.7623 38.5023	2.9159 18.1085	1.7825 20.3243	5,9882 17,7683	2.3144 18.2403	
400	3.2681 -0.4122	2.6544	0.9653 0.0211	_68.3077 _73.3581	42.4664 47.7633	14.4232 15.1337	20.3250	21.7030	21.8655	
800 BIASED CASE	0.4122	0.000	0.0211	13-3381	47.7633	12*1331	27.3129	23,2203	7 7 *0918	
×7 RAKED DATA BY SAMPLE ST2R										
50	81.2761	71.0930	43.3960	40.3490	39,2762	20.3311	60,9456	47.4610	21.2695	
100 400	66.5678 50.2068	53,2557 34,0251	26.6596 23.2212	80,6731 84,5035	0.6257 10.7395	3,2865 5,8094	21.2143 14.5921	15,2029 3,3799	3,7944 _10,2437	
800	46.4116	26.7202	11.5421	"85 . 5066	23 .7 086	3,7813	7.3590	~1. 9310	11.7206	
×4 RAKRD DATA BY SAMPLE SIZS										
50 1 J0	70.6965 53.3898	57.6538 43.5045	26.5811 19.4036	8.3320 68.2847	22.1949 2.9539	14.5839 8.2129	36.4610 15.4499	31.1395 9.5050	11.1017 8.7511	
400. 800.	47.3938 40.0275	34.5974 21.7419	22.6489 9.4810	72.1091 74.0936	12.1244 26.5986	5,5602 4,0235	13,5652 6,9843	_2.6844 -5.9180	11.4072 16.0150	
	40.0275	21.141	4010	14.0750	20.000	4.0203		3	10.0130	
UNBIASED CASE				PART II - 16.	KPONENTIAL—TYE	ν Ε' VARIABLES				
×7 RAKED DATA BY										
SAMPLE SIZE 50	30.9734	30.E219	18,2296	5.191J	0.9293	16.4981	14.2248	10,4409	18.0942	
100	9.9486	9.5417	7.8944	63.3927	38.3794	15,5653	16.857 €	15,3555	17.7314	
400 800	5.8796 0.6070	4.7602 0.1307	1.7130 0.0867	68.8586 68.2078	-44.1341 -48.4882	13.1318 12.8178	_17.3002 _20.3352	-17,1112 -18,3959	-18.7531 -17.1665	
×4 RAKED DATA BY SAMPLE SIZE				_	_					
50 100	10.8215 4.3724	10.3200 4.7549	8.1118 2.7422	-4,9458 -60,3050	-0.1392 -38.2203	_12.2000 _19.4866	0,6703 20,2154	4,4949 17,8034	15,4820 19,8590	
400. 800.	3.2221 1.4029	2.9617	_1.0779 -0.1658	¯65,7444 ¯62,7791	42.3684 46.9729	13,3866 13,6757	19,2563 19,4135	18,8287 17,5520	19,1187 17,1020	
BIASED CASE										
*7 RAKED DATA BY SAMPLE SIZS										
50	71.8517	69.0369	43.8727	_71.9420	60.4433	49.4973	63.5599	53.2207	46.1719	
100 400	65.4997 49.1219	53.6405 38.6396	29.4283 29.5335	15.5542 17.2178	25.0653 21.2627	3.2552 19.0808	30.1020 23.3482	24.1653 18.0932	1,5467 4,1554	
800	32.0277	16.8909	5,2030	21.1441	7.4373	6.0432	24,5550	9,0054	3,1899	
×4 RAKED DATA BY SAMPLE STZE 50	62.7472	59,3478	31.1049	47,8924	49.0686	43.7109	42,2690	43.9682	41.9225	
100	53.7795	45,2523	22.3463	1u 5031	20,8195	1.2840	26.7356	21.4924	1.6753	
400. 800.	44.4716 27.0033	37.2036 12.1416	28.7954 2.8674	17.6611 16.2339	16.3669 7.5887	17.4082 6.7783	19,8896 24,7736	16.1590 7.7126	3,7769 6,0839	

	••	DEGRBE OF DEPENDENCE BETWEEN RAKSO VARIABLES AND UNCONSTRAINED MEAN								
TTEM	COMP	LETE INDEPEN	DENCE	COM	IPLETE DEPENDI	INCE	PAI	RTTAL DEPENDE	VCE	
3154	RAKED MARGINALS		SIMPLE	RAKED MARGINALS		SIMPLE	RAKED MARGINALS		SIMPLE	
	UNRELATED	RELATED	RATIO ADJUSTMENT	UNRELATED	RELATED	RATIO ADJUSTMENT	UNRELATED	KELATED	RATIO ADJUSTMEN	
				PART I ~ '	JNIFORM—TYPE	' VARIABLES				
UNBIASED CASE ×7 RAK3D DATA BY SAMPLE SIZE										
50 100 400 800	4.2584 1.1178 0.1888 0.0023	-3.2377 -1.1180 -0.1419 -0.0833	3.0077 70.0340 0.0642 0.0146	1.9602 11.6578 5.3461 4.3310	2.7677 6.2042 3.8089 3.1360	0.6933 2.4682 1.2989 1.1943	3.5770 3.4181 1.8609 1.6969	-1.9820 3.2356 2.0898 1.6315	0.2331 3.0872 2.0965 1.5476	
×4 RAKED DATA BY SAMPLE SIZE	2.0834	-1.7794	0.8728	2 74.54	0.1926	0.0202	0,2698	0,6592		
50 100. 400. 800.	0.7030 0.1308 0.0056	1.7794 -0.5772 -0.0787 -0.0336	0.0728 70.4552 0.0085 0.0132	0.7141 9.9447 4.5554 3.9106	5,9930 3,4106 3,1270	0.0383 2.7057 1.1886 1.0936	0.2698 3.3427 1.8225 1.5848	3,2951 2,0561 1,5552	0.4302 3.0191 1.8972 1.4145	
BIASED CASE										
**7 RAKED DATA BY SAMPLE STZE 50. 100. 400. 800.	-14.8566 -8.4424 -3.4000 -2.4766	12.4916 6.4990 2.2545 1.4718	7.9128 3.4616 1.4175 0.5782	32,2359 58,3432 49,3061 47,7286	21.3223 32.9391 32.0629 32.2947	12.6171 16.2743 14.4902 14.9089	13.0204 25.9356 25.6897 25.5622	14.3031 22.6654 22.7820 22.4281	16.4983 19.9401 18.8001 18.1616	
×4 RAKBO DATA BY SAMPLE ST28 50 100 400 800	13.3055 7.1648 73.3727 2.1538	10.6614 5.5727 2.3358 1.1005	5.1257 2.5859 1.3586 0.3765	43.0031 54.7180 46.8456 45.5062	28.2065 32.5105 31.1288 31.4195	12.9587 16.0415 13.7750 14.2093	20.0537 25.2259 24.6614 24.6054	18.2096 22.3668 21.9247 21.8209	17.3212 19.7114 18.0824 17.6555	
	PART II - 'EXPONENTIAL-TIPE' VARIABLES									
UNBTASED CASE										
×7 RAK5D DATA BY SAMPLE ST25 50 100 400 800	-9.2866 -2.0639 -0.4132 0.0625	"9.0605 "2.0296 "0.25°1 "0.0398	5.8907 1.9045 0.0343 0.0487	"1.7770 17.1430 8.1043 6.0102	"3,3631 10,9978 6,0205 5,0138	"4.6363 4.3400 1.7371 1.4463	"6.5696 5.1674 2.5781 2.0524	-4,5825 4,8057 2,7666 2,0275	4.2997 5.0569 2.6559 1.7882	
×4 RAKSD DATA BY SAMPLE SIZE 50. 100. 400. 800.	-3.3816 -1.0127 -0.2198 0.1109	-3.1153 -0.8414 -0.1368 0.0821	2.7720 0.7326 0.0234 0.0569	2.0117 16.5198 7.9691 5.7855	-0.5084 11.0771 5.9899 5.0899	-3.0988 5.0001 1.6850 1.4438	0.2860 5.8538 2.7203 1.9778	-1.4830 5.4931 2.8938 2.0231	-3.7708 5.4451 2.5137 1.7107	
BIASED CASE										
x7 RAKED DAT4 BY SAMPLB SIZE 50 130 400 800	25.6297 14.2732 5.1939 2.8133	23,5760 11,3964 3,6185 1,3420	14.7632 6.7328 2.8264 0.2904	50.7664 92.0979 79.8462 77.6003	33,2929 53,9695 52,5672 54,2336	13.9117 28.1756 23.9549 24.9498	23.6881 45.6994 44.1076 43.0024	24.5806 38.9207 38.1280 37.8244	22.6809 34.7398 30.7860 30.4355	
4 RAKSD DATA BY SAMPLE STZ% 50 100 400. 800.	22.7273 12.2139 4.8420 2.3075	21.0292 10.0385 3.6375 0.9125	10.9359 5.0963 2.7473 0.0556	68.1091 87.8534 76.5852 74.0982	42.2231 53.6507 51.2524 52.4827	14.8229 27.7164 22.8141 23.7114	35.1064 43.6505 42.4497 41.2995	28.5877 37.5918 36.7123 36.5894	22.4109 33.9814 29.4343 29.4638	