

# **Simulation study to check the performance of various unequal probability sampling estimators**

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## **ABSTRACT**

A large number of unequal probability sampling estimators are available in literature. Theoretical comparison of available unequal probability sampling estimators is a hard task, so survey statisticians have conducted empirical studies to discuss the performances of various estimators. These empirical comparisons were based on limited number of populations available in the literature on survey sampling. The aim of this paper is to perform a simulation study on various popular unequal probability sampling estimators. This simulation study has been conducted by generating populations with given correlation structure from the Bi-Variate Normal Distribution. The study attempt to obtain a minimum variance estimator in unequal probability sampling for population with specific correlation structure.

**KEY WORDS :** Unequal Probability Sampling, Bi-variate Normal Data, Simulation

## **1. INTRODUCTION**

Unequal probability sampling has been a popular method of sample selection for estimation of population characteristics. The paper deals with comprehensive comparison of some procedures of unequal probability sampling by Simulation. The available procedures are not mathematical comparable because of complex nature of equations and conditions. Lot of empirical comparisons has been done from time to time to obtain a minimum variance estimator or selection procedure for unequal

probability sampling. These empirical comparisons have been limited to a specific set of populations and no correlation structures have been specified in these empirical comparison. In this article we have carried out a simulation study to search for an optimum selection procedure that can be used with the Horvitz – Thompson (1952) estimator. This simulation study has been carried out by generating random data from bi-variate normal population with specific correlation structure. The methods that are compared are given in the following section.

## 2. METHODS USED FOR COMPARISON

In this section the methods used for comparison are given. These methods are selected as they have been widely used in many empirical studies by number of survey statisticians. The methods are listed below:

### 2.1: Horvitz-Thompson Estimator (1952)

The Horvitz – Thompson (1952) estimator is given as:

$$\text{Estimate : } y'_{HT} = \sum_{i \in S} \frac{Y_i}{\pi_i} \quad (2.1)$$

$$\text{Variance: } Var(y'_{HT}) = \frac{1}{2} \sum_{\substack{i=1 \\ j \neq i}}^N \sum_{j=1}^N (\pi_i \pi_j - \pi_{ij}) \left( \frac{y_i}{\pi_i} - \frac{y_j}{\pi_j} \right)^2 \quad (2.2)$$

The Horvitz – Thompson (1952) estimator has been used under following selection procedures:

### 2.2.1: Yates Grundy (1953) draw by draw Procedure:

$$\pi_i = p_i \left[ 1 + \sum_{j=1}^N \frac{p_j}{1-p_j} - \frac{p_i}{1-p_i} \right] \quad (2.3)$$

$$\pi_{ij} = p_i p_j \left[ \frac{1}{1-p_i} + \frac{1}{1-p_j} \right] \quad (2.4)$$

### 2.2.2: Brewer (1963) Procedure:

$$\pi_i = 2 p_i$$

$$\pi_{ij} = \frac{2 p_i p_j}{k} \left[ \frac{1}{1-2 p_i} + \frac{1}{1-2 p_j} \right] \text{ with } k = 1 + \sum_{j=1}^N \frac{p_j}{1-2 p_j} \quad (2.5)$$

### 2.2.3: Yates – Grundy Rejective Procedure (1953)

$$\pi_i = \frac{2 p_i (1 - p_i)}{1 - \sum_{j=1}^N p_j^2} \quad (2.6)$$

$$\pi_{ij} = \frac{2 p_i p_j}{1 - \sum_{j=1}^N p_j^2} \quad (2.7)$$

### 2.2.4: Shahbaz and Hanif Procedure (2003)

$$\pi_i = \frac{p_i}{d} \left[ \frac{1}{1 - p_i} + \sum_{j=1}^N \frac{p_j}{(1 - p_j)(1 - 2 p_j)} \right]; \quad d = \sum_{i=1}^N \frac{p_i}{1 - 2 p_i} \quad (2.8)$$

$$\pi_{ij} = \frac{p_i p_j}{d} \left[ \frac{1}{(1 - p_i)(1 - 2 p_i)} + \frac{1}{(1 - p_j)(1 - 2 p_j)} \right] \quad (2.9)$$

## 2.2 Murthy (1957) Estimator:

$$\text{Estimate : } t_{\text{symm}} = \frac{1}{2 - p_i - p_j} \left[ \frac{y_i}{p_i} (1 - p_j) + \frac{y_j}{p_j} (1 - p_i) \right] \quad (2.10)$$

$$\text{Var}(t_{\text{symm}}) = \frac{1}{2} \sum_{i=1}^N \sum_{\substack{j=1 \\ j \neq i}}^N \frac{p_i p_j (1 - p_i - p_j)}{2 - p_i - p_j} \cdot \left( \frac{Y_i}{p_i} - \frac{Y_j}{p_j} \right)^2 \quad (2.11)$$

The simulation study is given in the following section.

### 3. SIMULATION STUDY

In this section the results of simulation study have been given. The study has been carried out by drawing random data from bi-variate normal distribution. The random data has been drawn by using various values of correlation coefficient. After drawing the random data; the variance of Horvitz-Thompson Estimator (1952) has been computed under various selection procedures alongside variance of Murthy estimator. The procedure has been replicated various number of times. After this replication the average variance of each method has been computed for various values of correlation coefficient. The results of the study have been given in the table 3.1. Following abbreviations have been used in table 3.1

**Vyg** Variance of Yates Grundy (1953) draw by draw procedure

**Vr** Variance of Yates Grundy (1953) rejective procedure

**Vb** Variance of Brower (1963) procedure

**Vs** Variance of Shahbaz and Hanif (2003)

**Vm** Variance of Murthy (1957) estimator

**Raho** Population Correlation Coefficient

**Table 3.1: Average Variances of Various Estimators for different number of replicates and different Correlation Structure**

Replicate		Rho									
		0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1
10	Vyg	367814	314255	261634	208861	203721	222588	155648	146938	128165	127014
	Vb	373242	316477	263982	209528	203734	221543	154544	145238	124730	121554
	Vr	363146	312431	259699	208400	203826	223634	156703	148521	131255	131881
	Vm	371720	315654	262921	209056	203417	220460	154251	144740	124219	121104
25	Vs	378142	318557	266149	210238	203865	220761	153684	143848	121835	116948
	Vyg	322869	257688	230817	217205	221015	188407	157279	158859	112280	121020
	Vb	326943	260214	232414	217912	219976	187447	155622	156558	108731	116633
	Vr	319415	255588	229532	216683	222033	189344	158841	160969	115476	124943
50	Vm	325910	259669	231733	217440	219487	187132	155044	155907	108379	115901
	Vs	330631	262540	233922	218645	219196	186703	154299	154644	105751	112912
	Vyg	341313	276598	254699	257192	187098	181670	158427	155054	123453	105956
	Vb	346060	278795	255863	258298	187003	180741	156402	152696	120007	101695
100	Vr	337277	274806	253799	256336	187280	182592	160295	157224	126555	109763
	Vm	344594	278257	255244	257662	186648	180232	155995	152052	119548	101196
	Vs	350346	280822	256992	259379	187024	180041	154747	150742	117105	98089
	Vyg	289818	271476	257316	241046	198177	172609	164012	153776	130060	110205
250	Vb	293382	274201	259036	241583	198224	171567	162354	150788	126860	105909
	Vr	286812	269199	255921	240684	198240	173621	165557	156489	132939	114038
	Vm	292543	273500	258304	240934	197638	171117	161914	150374	126232	105375
	Vs	296622	276695	260648	242171	198373	170759	161005	148297	124180	102277
500	Vyg	300222	282311	258937	228193	206228	184452	166560	148726	128352	108019
	Vb	303650	284474	260428	229203	206066	183428	164680	146109	124901	103788
	Vr	297332	280524	257743	227416	206472	185448	168303	151109	131458	111800
	Vm	302834	283813	259765	228570	205578	182983	164203	145622	124408	103262
1000	Vs	306763	286476	261846	230198	206035	182645	163146	143932	121995	100203
	Vyg	292662	274989	255172	225084	215332	194098	165507	150412	125342	104594
	Vb	295813	277466	256786	225681	215180	193002	163766	147704	121854	100526
	Vr	290013	272928	253867	224666	215570	195159	167131	152871	128482	108229
2500	Vm	295015	276743	256091	225111	214647	192501	163251	147249	121376	100004
	Vs	298685	279742	258313	226312	215158	192156	162348	145446	118919	97079
	Vyg	305239	276436	253448	230203	210368	187332	164127	145648	125993	102516
	Vb	308630	278862	254996	230899	210241	186398	162338	143169	122559	98435
5000	Vr	302379	274421	252205	229700	210587	188253	165790	147909	129083	106162
	Vm	307752	278101	254375	230312	209661	185929	161869	142693	122072	97955
	Vs	311712	281097	256463	231618	210241	185691	160884	141109	119671	94978
	Vyg	297927	277918	255518	235577	210328	188142	166078	145403	125413	104538
10000	Vb	301263	280361	257121	236327	210231	187211	164328	142804	122030	100419
	Vr	295116	275888	254226	235032	210520	189059	167708	147772	128460	108217
	Vm	300398	279597	256473	235726	209683	186718	163853	142317	121536	99924
	Vs	304298	282613	258636	237094	210256	186507	162907	140640	119183	96930
25000	Vyg	304084	276182	257168	229185	208605	188005	167797	144873	125584	105463
	Vb	307459	278567	258784	229870	208435	187049	166012	142305	122196	101291
	Vr	301237	274203	255865	228694	208859	188944	169458	147211	128633	109190
	Vm	306574	277840	258116	229279	207890	186544	165516	141831	121713	100794
50000	Vs	310529	280766	260309	230579	208397	186323	164561	140169	119345	97756
	Vyg	299232	275627	254391	234400	208839	187887	167005	145727	125005	104743
	Vb	302605	278038	255938	235090	208680	186939	165229	143134	121637	100623
	Vr	296388	273625	253146	233905	209085	188821	168658	148090	128037	108425
100000	Vm	301757	277287	255269	234473	208128	186428	164747	142662	121155	100118
	Vs	305671	280261	257405	235804	208651	186220	163786	140977	118803	97132
	Vyg	299785	277551	253927	231460	209898	187516	165986	145536	124950	106295
	Vb	303119	280010	255512	232170	209748	186512	164196	142944	121572	102125
250000	Vr	296975	275509	252649	230949	210136	188499	167651	147897	127991	110021
	Vm	302269	279261	254841	231566	209204	186012	163717	142470	121092	101607
	Vs	306150	282274	257011	232902	209728	185744	162740	140788	118730	98593
	Vyg	300857	277028	254536	232143	209521	187660	166735	145375	125122	105320
500000	Vb	304207	279462	256116	232853	209379	186676	164940	142779	121739	101168
	Vr	298033	275008	253262	231631	209752	188625	168405	147739	128168	109029
	Vm	303354	278714	255451	232253	208833	186173	164456	142306	121258	100662
	Vs	307254	281704	257610	233586	209365	185925	163480	140620	118893	97651

\* More shaded circle represent the larger variance

#### 4. CONCLUSIONS

The results of the simulation study have been given in section 3 of the article. The table 3.1 contains the average variance of various selection procedures under different correlation structure. The table has been constructed by using various numbers of replicates from the bivariate normal distribution. From the table we can see that Yates–Grundy (1953) rejective procedure outperform other selection procedures involved in the study for populations having correlation between 0.1 to 0.5. The procedure given by Shahbaz–Hanif (2003) outperform other procedures involved in the study for populations having correlation of 0.6 to 1.0, and this procedure is closely followed by the Murthy (1957) estimator. The other procedures do not perform batter.

From the table 3.1; we can, therefore, conclude that the Yates–Grundy (1953) rejective procedure is a batter choice for estimation of population total by using the Horvitz–Thompson (1952) estimator for populations having a correlation coefficient of below 0.5. The Shahbaz – Hanif (2003) procedure is better for populations having high correlation; that is correlation of greater than or equal to 0.6.

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