

# STA304/1003 H1 F - Summer 2014: Surveys, Sampling, and Observational Data

## Lecture 4 - Part II: Stratified Random Sampling

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## When should you use a SRS?

Simple Random Samples are the easy to design and analyze, but may not be appropriate in some cases.

### Use a SRS when:

- ▶ Little/no extra information is available about characteristics in the population
- ▶ Data users insist on SRS formulas: averaging sample values
- ▶ Main interest is multivariate relationships (regression equations) for the population: easier to perform and interpret for SRSs

### Do NOT use a SRS when:

- ▶ A controlled experiment is appropriate (not a survey sample)
- ▶ List of observation units in population is not available or too expensive/time consuming to take SRS
- ▶ You have additional information about population characteristics that can improve survey design / cost effective design

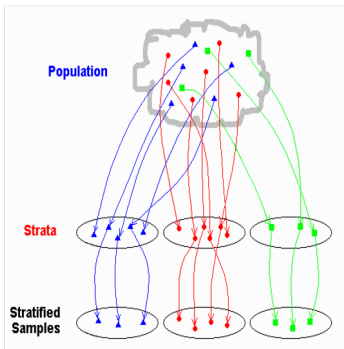
# Stratified Random Sampling

Recall that in **Stratified Random Sampling (STRS)**:

1. Population split into  $L$  distinct **strata** / groups:

Strata should partition the population (should **not overlap** and should comprise of whole population so that each sampling unit belongs to exactly one stratum).

2. **Take independent probability samples (SRS) from each stratum**
3. **Pool information** to get overall population parameters



## Why choose a STRS?

- ▶ Can obtain more representative sample than SRS
- ▶ Elements homogeneous within strata:  
Smaller variances / more precise estimates  
→ narrower CIs      *Think about "age groups"*
- ▶ Cost most likely lower, more convenient, easier to administer than SRS:  
Can use different sampling procedures for different strata
- ▶ May be interested in estimates within subpopulations with known precision:  
Choose subpopulations as strata, sample according to population proportions or depending on precision

## Examples: How would you stratify in each case?

- (1) A study about blood pressure *ages/gender/BMI/...*
- (2) A study about concentration of plants in an area *rainfall/temperature/size*
- (3) Political Survey *minority groups ...*
- (4) Absences of Primary School Children Example *age/grade ...*
- (5) A study on salaries of university instructors *faculty! ...*

## Theory and Notation for STRS

- ▶ Divide population of size  $N$  into  $L$  strata with  $N_i$  sampling units in stratum  $i$
- ▶  $N_1, N_2, \dots, N_{L-1}, N_L$  population sizes known and  $N = \sum_{i=1}^L N_i$
- ▶ Take SRS of size  $n_i$  from each stratum, denoted  $\mathcal{S}_i$
- ▶ Total sample size:  $n = \sum_{i=1}^L n_i$
- ▶  $i = 1, \dots, L$  : index for strata
- ▶  $j = 1, \dots, N_i$  : index for elements within stratum  $i$

Population parameters are:

- ▶  $y_{ij}$  : variable/measurement value of  $j$ th unit in stratum  $i$
- ▶  $\tau_i = \sum_{j=1}^{N_i} y_{ij}$  : Population total in stratum  $i$
- ▶  $\tau = \sum_{i=1}^L \tau_i$  : Population total (overall)
- ▶  $\bar{y}_{iU} = \frac{1}{N_i} \sum_{j=1}^{N_i} y_{ij}$  : Population mean in stratum  $i$
- ▶  $\bar{y}_U = \frac{\tau}{N} = \frac{\sum_{i=1}^L \sum_{j=1}^{N_i} y_{ij}}{N}$  : Population mean (overall)
- ▶  $S_i^2 = \frac{1}{N_i-1} \sum_{j=1}^{N_i} (y_{ij} - \bar{y}_{iU})^2$  : Population variance within stratum  $i$
- ▶  $S^2 = \frac{1}{N-1} \sum_{i=1}^L \sum_{j=1}^{N_i} (y_{ij} - \bar{y}_U)^2$  : Population variance (overall) - may not be useful!

## Sample Quantities / Estimators

Use SRS estimators within each stratum to obtain:

- ▶  $\bar{y}_i = \frac{1}{n_i} \sum_{j \in \mathcal{S}_i} y_{ij}$  : estimates  $\bar{y}_{iU}$
- ▶  $\hat{\tau}_i = \frac{N_i}{n_i} \sum_{j \in \mathcal{S}_i} y_{ij} = N_i \bar{y}_i$  : estimates  $\tau_i$
- ▶  $s_i^2 = \frac{1}{n_i - 1} \sum_{j \in \mathcal{S}_i} (y_{ij} - \bar{y}_i)^2$  : estimates  $S_i^2$
- ▶  $\hat{\tau}_{st} = \sum_{i=1}^L \hat{\tau}_i = \sum_{i=1}^L N_i \bar{y}_i$  : estimates  $\tau$
- ▶  $\bar{y}_{st} = \frac{\hat{\tau}_{st}}{N} = \sum_{i=1}^L \frac{N_i}{N} \bar{y}_i$  : estimates  $\bar{y}_U$

weight:  $\frac{N_i}{N}$

↪ Weighted average of sample stratum averages, weights are proportions of population units in each stratum.



- Must know sizes or relative sizes of strata to use STRS

## Properties of Estimators

- Unbiasedness:

$\bar{y}_{st}$  is unbiased for  $\bar{y}_U$  and  $\hat{\tau}_{st}$  is unbiased for  $\tau$

- Variances:

$$V(\bar{y}_{st}) = \sum_{i=1}^L \left(1 - \frac{n_i}{N_i}\right) \left(\frac{N_i}{N}\right)^2 \frac{S_i^2}{n_i}$$

$$V(\hat{\tau}_{st}) = \sum_{i=1}^L \left(1 - \frac{n_i}{N_i}\right) N_i^2 \frac{S_i^2}{n_i}$$

- Standard Errors:

In order to estimate variances, we need to sample at least 2 units from each stratum. O.W. 1 unit  $\Rightarrow$  Variance = 0  
 $(1 - \frac{1}{1}) = 0 \checkmark$

$$SE(\bar{y}_{st}) = \sqrt{\sum_{i=1}^L \left(1 - \frac{n_i}{N_i}\right) \left(\frac{N_i}{N}\right)^2 \frac{s_i^2}{n_i}}$$

$$SE(\hat{\tau}_{st}) = \sqrt{\sum_{i=1}^L \left(1 - \frac{n_i}{N_i}\right) N_i^2 \frac{s_i^2}{n_i}}$$



## Proofs of Properties

Remember:

- Properties of SRS estimators
- Properties of expectations and variances
- Independence when sampling from strata

Recall:  $\bar{y}_{st} = \sum_{i=1}^L \frac{N_i}{N} \bar{y}_i$

Since an SRS is taken from each stratum  $i$ ,  $E(\bar{y}_i) = \bar{y}_{iu}$

so,  $E(\bar{y}_{st}) = \sum_{i=1}^L \frac{N_i}{N} E(\bar{y}_i) = \frac{1}{N} \left( \sum_{i=1}^L N_i \bar{y}_{iu} \right) = \frac{T}{N} = \bar{y}_u$

Since an SRS is taken from each stratum  $i$ ,  $\text{Var}(\bar{y}_i) = \left(1 - \frac{n_i}{N_i}\right) \frac{S_i^2}{n_i}$

So  $\text{Var}(\bar{y}_{st}) = \text{Var}\left(\sum_{i=1}^L \frac{N_i}{N} \bar{y}_i\right) = \sum_{i=1}^L \text{Var}\left(\frac{N_i}{N} \bar{y}_i\right)$  since each stratum indep from each others

$$= \sum_{i=1}^L \frac{N_i^2}{N^2} \text{Var}(\bar{y}_i) = \sum_{i=1}^L \left(\frac{N_i}{N}\right)^2 \left(1 - \frac{n_i}{N_i}\right) \frac{S_i^2}{n_i}$$

Since  $\hat{T}_{st} = N \bar{y}_{st}$ ,  $E(\hat{T}_{st}) = N E(\bar{y}_{st}) = N \bar{y}_u = T$

$$\text{Var}(\hat{T}_{st}) = N^2 \text{Var}(\bar{y}_{st}) = \sum_{i=1}^L N_i^2 \frac{S_i^2}{n_i} \left(1 - \frac{n_i}{N_i}\right)$$

## Confidence Intervals

If either:

- (1) Sample sizes within each stratum are large OR
- (2) Large number of strata

Then,

An approximate  $100(1 - \alpha)\%$  CI for the population mean,  $\bar{y}_U$  is:

$$\bar{y}_{st} \pm z_{\alpha/2} SE(\bar{y}_{st})$$

An approximate  $100(1 - \alpha)\%$  CI for the population total,  $\tau$  is:

$$\hat{\tau}_{st} \pm z_{\alpha/2} SE(\hat{\tau}_{st})$$

\* Note: Some software use  $t_{n-L}$  critical values rather than standard normal \*

## Installing Sampling Contributed Package in 'R'

1. Open R
2. Be sure you are connected to the internet
3. At the top of the R window click on **Packages**
4. A list will open, click on **Install Packages**
5. A list of mirror sites appears. Select **Canada (ON)**, and click **OK**
6. Another list will open, click on **Sampling** and then click **OK**
7. A lot of information will appear on the screen, but at the end you will get the R prompt >
8. Again click **Packages**, then click **Load Package**, select **Sampling** and click **OK**

## Example: Using R for Stratified Sampling

Groups A,B,C,D and one variable (response)

> attach(strsex)

Read the data into R:

```
> strsex<-read.csv("strsex.csv")
> strsex
```

	response	group
1	8.8	A
2	10.6	A
3	10.6	A
4	7.6	A
5	7.7	A
6	10.0	A
. . .		
75	8.3	D
76	12.3	D
77	9.4	D
78	7.9	D
79	6.9	D
80	11.2	D

Find population mean and total:

```
> mean(strsex$response)
[1] 9.8325
> sum(strsex$response)
[1] 786.6
> sum(strsex$response)/length(strsex$response)
[1] 9.8325
```

## Take a STRS:

```
> strs.sample<-strata(strsex,c("group"),size=c(3,4,5,6),method=c("srswor"))
```

```
> strs.sample
```

	group	ID_unit	Prob	Stratum
1	A	12	0.15	1
2	A	14	0.15	1
3	A	19	0.15	1
4	B	23	0.20	2
5	B	30	0.20	2
6	B	33	0.20	2
7	B	36	0.20	2
8	C	43	0.25	3
9	C	46	0.25	3
10	C	48	0.25	3
11	C	49	0.25	3
12	C	51	0.25	3
13	D	64	0.30	4
14	D	67	0.30	4
15	D	68	0.30	4
16	D	69	0.30	4
17	D	75	0.30	4
18	D	77	0.30	4

## Look at STRS data:

```
> strs.sample.data<-getdata(strsex,strs.sample)
```

```
> strs.sample.data
```

	response	group	ID_unit	Prob	Stratum
1	9.3	A	12	0.15	1
2	9.4	A	14	0.15	1
3	13.2	A	19	0.15	1
4	11.1	B	23	0.20	2
5	8.4	B	30	0.20	2
6	10.2	B	33	0.20	2
7	10.1	B	36	0.20	2
8	10.5	C	43	0.25	3
9	7.7	C	46	0.25	3
10	7.9	C	48	0.25	3
11	10.3	C	49	0.25	3
12	7.5	C	51	0.25	3
13	7.5	D	64	0.30	4
14	11.8	D	67	0.30	4
15	6.1	D	68	0.30	4
16	9.2	D	69	0.30	4
17	8.3	D	75	0.30	4
18	9.4	D	77	0.30	4

Calculate  $N_i$ ,  $n_i$ ,  $\bar{y}_i$ ,  $s_i^2$  for each stratum:

```
> Ni<-tapply(strsex$response, strsex$group, length)
> ni<-tapply(strs.sample.data$response, strs.sample.data$group, length)
> ssqi<-tapply(strs.sample.data$response, strs.sample.data$group, var)
> ybari<-tapply(strs.sample.data$response, strs.sample.data$group, mean)
> Ni
  A  B  C  D
20 20 20 20
> ni
A B C D
3 4 5 6
> ssqi
      A      B      C      D
4.943333 1.270000 2.212000 3.741667
> ybari
      A      B      C      D
10.633333  9.950000  8.780000  8.716667
```

Population size:

```
> N = length(strsex$response)
> N
[1] 80
```

Calculate  $\bar{y}_{st}$ :

```
> ybar.st<-sum(Ni*ybari)/N  
> ybar.st  
[1] 9.52
```

Calculate  $\hat{V}(\bar{y}_{st})$ :

```
> var.ybar.st<-sum(Ni^2*(1-ni/Ni)*ssqi/ni)/N^2  
> var.ybar.st  
[1] 0.1514337
```

Calculate  $\hat{\tau}_{st}$ :

```
> N*ybar.st  
[1] 761.6
```

Calculate  $\hat{V}(\hat{\tau}_{st})$ :

```
> N^2*var.ybar.st  
[1] 969.1755
```



## Example: Confidence Intervals

$$\bar{y}_{st} \pm 1.96 \sqrt{\hat{\text{Var}}(\bar{y}_{st})}$$

$$= 9.52 \pm 1.96 \sqrt{0.1514} = (8.7574, 10.2826)$$

- a) Use the R output and data to find a 95% CI the population mean and a 95% CI for the population total (assuming the required assumptions are met).

$$N=80$$

$$\bar{y}_{st}=9.52$$

$$\hat{\text{Var}}(\bar{y}_{st})=0.1514$$

$$\text{true value } \bar{y}_U = 9.8325$$

contained in CI

good ✓

Since  $\hat{t}_{st} = N \bar{y}_{st}$

$$N=80$$

95% CI for  $\hat{t}$  is  $(N \times 8.7574, N \times 10.2826)$

$$= (700.592, 822.608)$$

- b) Use the R output to find a 95% CI for the mean of group D (assuming the required assumptions are met).

Actually using R: 95% CI for  $\bar{y}_D$ :  $n_D=6$   
from group D  $N_D=20$

$$\bar{y}_D \pm 1.96 \sqrt{\left(1 - \frac{n_D}{N_D}\right) \frac{s_D^2}{n_D}} = 8.7167 \pm 1.96 \sqrt{\left(1 - \frac{6}{20}\right) \frac{3.7417}{6}} = (7.4217, 10.0117)$$

## Stratified Sampling for Proportions

Recall that proportions are simply means of indicator variables.

Use:  $\hat{p}_i = \bar{y}_i$  and  $s_i^2 = \frac{n_i}{n_i - 1} \hat{p}_i(1 - \hat{p}_i)$ .

$$\hat{p}_{st} = \sum_{i=1}^L \frac{N_i}{N} \hat{p}_i$$

$$\hat{V}(\hat{p}_{st}) = \sum_{i=1}^L \left(1 - \frac{n_i}{N_i}\right) \left(\frac{N_i}{N}\right)^2 \frac{\hat{p}_i(1 - \hat{p}_i)}{n_i - 1}$$

An approximate  $100(1 - \alpha)\%$  CI for the proportion,  $p$  is:

$$\hat{p}_{st} \pm z_{\alpha/2} SE(\hat{p}_{st})$$

## Estimating Total Number of Population Units with a Characteristic

$$\hat{\tau}_{st} = \sum_{i=1}^L N_i \hat{p}_i$$

i.e. the estimated total number of population units with the characteristic = sum of the estimated totals in each stratum

$$\hat{V}(\hat{\tau}_{st}) = N^2 \hat{V}(\hat{p}_{st})$$

An approximate  $100(1 - \alpha)\%$  CI for the population total,  $\tau$  is:

$$\hat{\tau}_{st} \pm z_{\alpha/2} SE(\hat{\tau}_{st})$$

## Example: Television Advertising

An advertising firm is interested in estimating the proportion of households in a certain county that watch TV show 'X', in order to target their advertising more efficiently. The county has two towns, A and B, and a rural area - Town A is built around a factory and most households contain factory workers with school-age children, while Town B contains mostly elderly residents with few children at home.

Location	Population Size	Sample Size	# of households viewing show 'X'
Town A	155	20	16
Town B	62	8	2
Rural	93	12	6

- Discuss the merits of using STRS in this case.
- Estimate the proportion of households in this county that view 'X' and place a bound on the error of the estimation (based on 95% confidence).

## Sampling Weights

$\pi_{ij} = \frac{n_i}{N_i}$ , so the sampling weights are:

$$w_{ij} = \frac{1}{\pi_{ij}} = \frac{N_i}{n_i}$$

- ▶ sampling weight interpreted as the number of units in the population represented by the sample member  $y_{ij}$ : each sampled unit in stratum  $i$  represents itself +  $\left(\frac{N_i}{n_i} - 1\right)$  other units in stratum  $i$  that were not selected in the sample
- ▶ sum of the weights is  $N$

▶

$$\hat{\tau}_{st} = \sum_{i=1}^L \sum_{j \in \mathcal{S}_i} w_{ij} y_{ij} \quad \text{and} \quad \bar{y}_{st} = \frac{\sum_{i=1}^L \sum_{j \in \mathcal{S}_i} w_{ij} y_{ij}}{\sum_{i=1}^L \sum_{j \in \mathcal{S}_i} w_{ij}}$$

- ▶ STRS is self-weighting if the sampling fraction  $\frac{n_i}{N_i}$  is the same for each stratum (i.e. sampling weight is  $\frac{N}{n}$  like for SRS. But variance depends on stratification - weights do not tell you the stratum membership of observations)