

# **NON-RESPONSE IN SAMPLE SURVEYS**

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## ❖ Missing Data Problems in Sample Surveys

- Missing data is a common problem and challenge for analysts.
- Non-response error arises from failure to obtain the responses from the respondents (sample units) during the survey.



## ❖ Causes of Missing Data Problems in Sample Surveys

- There are many reasons why data could be missing, including:



- Unavailable (No Contact)
- Unwillingness (Refusal)
- Unable (Not able)

- A sensor failed.
- Someone purposefully turned off recording equipment.
- There was a power cut.
- The method of data capture was changed.

- An internet connection was lost.
- A network went down.
- A hard drive became corrupt.
- A data transfer was cut short.



## ❖ **Methods to Handel the Non-Response Problems**

- To deal with the problem of non-response following techniques are frequently being used by survey practitioners

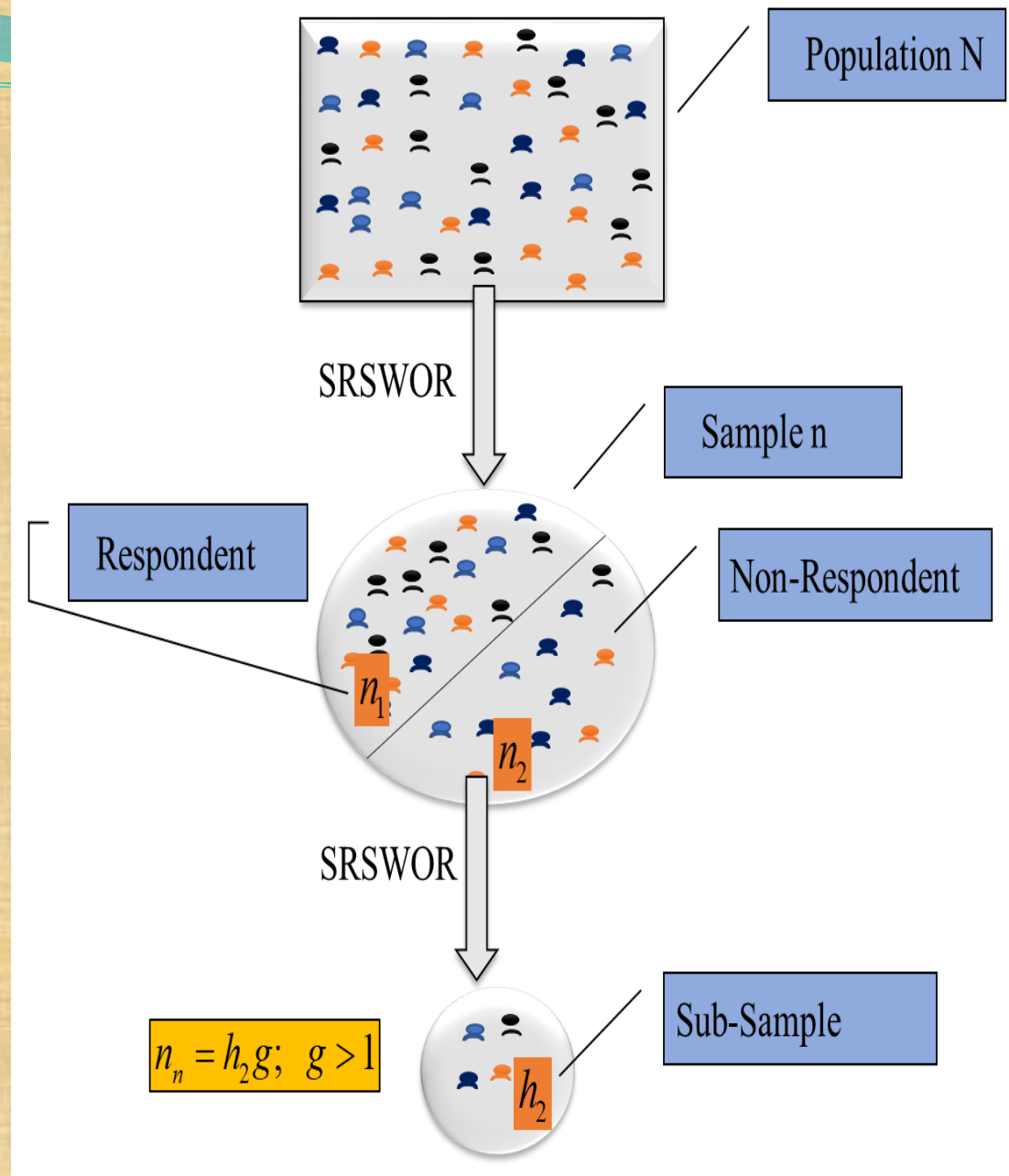
### ❖ **Hansen and Hurwitz (1946) Technique**

- Hansen and Hurwitz (1946) suggested a sub-sampling of non-respondents technique to deal with the problems of non-response which often occur in mail surveys. This technique is applicable for human surveys.

The Hansen and Hurwitz (1946) technique may be summarized in the following steps:

**STEP 1:** Select a sample of respondents from the population and seek their responses through postal mail, email or online survey etc. by a fixed deadline.

**STEP 2:** Once the deadline is over, identify the non-respondents.



**STEP 3:** Select a sub-sample from the non-respondents and seek their responses through more engaging mode of surveying such as personal interview by trained interviewers.

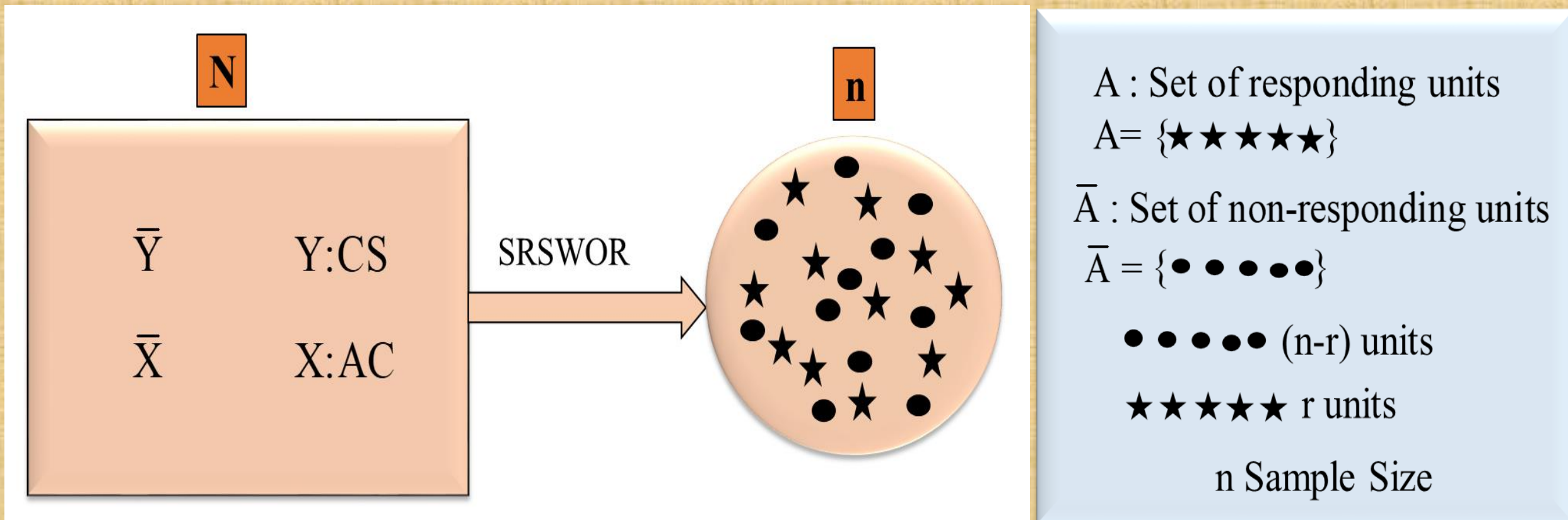
**STEP 4:** Combine the data of both the rounds of survey to develop an estimate for the population parameter.

The unbiased estimator of population mean  $\bar{Y}$  is given as

$$\bar{y}_{HH} = \frac{n_1 \bar{y}_1 + n_2 \bar{y}_{h_2}}{n}$$

where  $\bar{y}_{h_2}$  denotes the mean of  $h_2$  observations from the sub-sample of non-responding units.

## ❖ Imputation Technique





## ❖ General Imputation Method

$$y_{.i} = \begin{cases} y_i & \text{if } i \in A \\ \hat{y}_i & \text{if } i \in \bar{A} \end{cases}$$

$\hat{y}_i$  denotes the imputed value for the  $i^{\text{th}}$   
non-responding unit

- The general point estimator of population mean takes the form

$$\bar{y}_s = \frac{1}{n} \left[ \sum_{i=1}^r y_i + \sum_{i=1}^{n-r} \hat{y}_i \right]$$

where the value  $\hat{y}_i$  is *the imputed value*



## ❖ Mean Method of Imputation

- In this method, no auxiliary information is used.
- The missing values are replaced with the mean of responding units of the study variable.
- Under this method the data after imputation becomes:

$$y_{.i} = \begin{cases} y_i & \text{if } i \in A \\ \bar{y}_r & \text{if } i \in \bar{A} \end{cases}$$

$\bar{y}_r$  is the response mean

- The point estimator of the population mean is :

$$\bar{y}_m = \frac{1}{n} \left[ \sum_{i=1}^r y_i + \sum_{i=1}^{n-r} \bar{y}_r \right] = \bar{y}_r$$

## ❖ Ratio Method of Imputation

- In the ratio method of imputation, we assume that imputation is carried out with the aid of an auxiliary variable  $x$ .
- The data after imputation becomes:

$$y_{.i} = \begin{cases} y_i & \text{if } i \in A \\ \hat{b} x_i & \text{if } i \in \bar{A} \end{cases}$$

$$\text{where } \hat{b} = \frac{\sum_{i=1}^r y_i}{\sum_{i=1}^r x_i} = \frac{\bar{y}_r}{\bar{x}_r}$$

- The point estimator of the population mean becomes

$$\bar{y}_{RAT} = \frac{1}{n} \left[ \sum_{i=1}^r y_i + \sum_{i=1}^{n-r} \hat{b} x_i \right] = \frac{\bar{y}_r}{\bar{x}_r} \bar{x}_n$$

$$\Rightarrow \bar{y}_{RAT} = \frac{\bar{y}_r}{\bar{x}_r} \bar{x}_n$$

## ❖ Hot Deck (HD) Method of Imputation

$$y_{.i} = \begin{cases} y_i & \text{if } i \in A \\ \bar{y}_{g(i)} & \text{if } i \in \bar{A} \end{cases}$$

where  $y_{g(i)}$  is the  $y$  value given by the donor unit  $g(i) \in A$ ,  
drawn at random (with replacement) from the  $r$  responding units.

- Under the HD method of imputation the point estimator of population mean

$$\bar{y}_{HD} = \frac{1}{n} \left[ \sum_{i=1}^r y_i + \sum_{i=1}^{n-r} y_{g(i)} \right]$$



## ❖ Nearest Neighbour (NN) Method of Imputation

- Under the NN method the data after imputation becomes

$$y_{.i} = \begin{cases} y_i & \text{if } i \in A \\ \bar{y}_{g(i)} & \text{if } i \in \bar{A} \end{cases}$$

where  $y_{g(i)}$  is the  $y$  value given by the donor unit  $g(i)$  such that  $\underset{g \in R_p}{\text{Min}} |x_g - x_i|$  occurs for  $g = g(i)$ .

If it results in more than one unit a donor is randomly selected from them.

- Under the NN method of imputation the point estimator of the population mean becomes

$$\bar{y}_{NN} = \frac{1}{n} \left[ \sum_{i=1}^r y_i + \sum_{i=1}^{n-r} y_{g(i)} \right]$$

## ❖ Regression Method of Imputation

$$y_{.i} = \begin{cases} y_i & \text{if } i \in A \\ \hat{y}_i & \text{if } i \in \bar{A} \end{cases}$$

$$\text{where } \hat{y}_i = \hat{a} + \hat{b}x_i$$

$$\hat{b} = \frac{s_{yx}(r)}{s_x^2(r)}; \quad \hat{a} = \bar{y}_r - \hat{b}\bar{x}_r$$

- Under the Regression method of Imputation the point estimator of the population mean

$$\bar{y}_{Reg} = \frac{1}{n} \left[ \sum_{i=1}^r y_i + \sum_{i=1}^{n-r} (\hat{a} + \hat{b}x_i) \right]$$

$$\Rightarrow \bar{y}_{reg} = \bar{y}_r + \hat{b}(\bar{x}_n - \bar{x}_r)$$

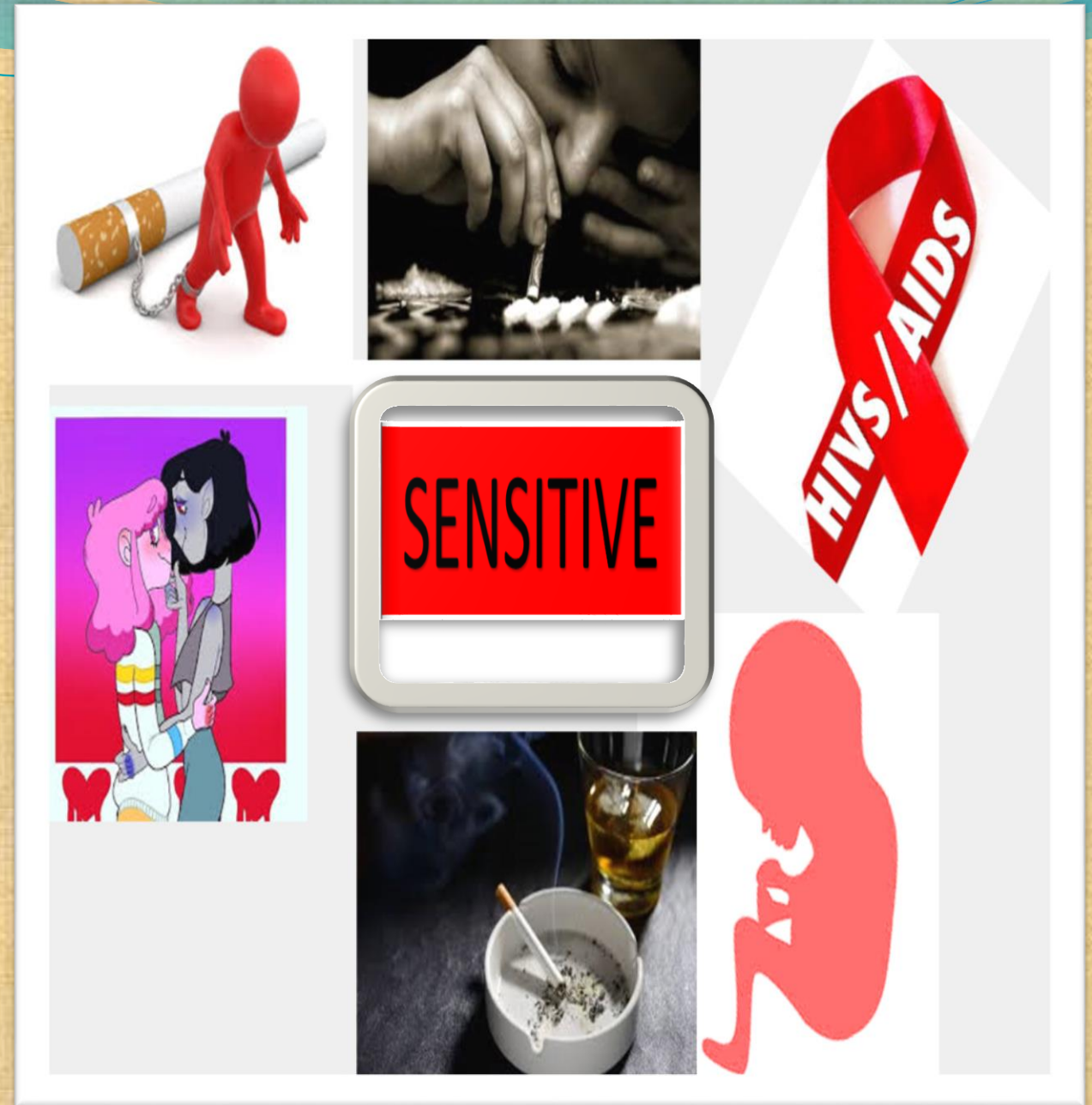
## ❖ Dealing with non-response arise due to sensitive issues

- Economists, psychologists, sociologists, managers, and policy makers have many reasons for asking personal questions.
- Biometric sample surveys need get ready-made information for future planning and policy implementations related to the subject matters of highly sensitive issues.





- Highly sensitive issues such as sexual behavior, domestic violence, tax offender, HIV infection status, drug addiction, extra marital affair etc.
- Actual answers of these questions are hidden or misguided by people.
- Data obtained are definitely open to error if surveys are conducted through classical methods.



# ❖ Randomized Response Technique

- Introduced by Warner (1965)
- When study under characteristics is sensitive in nature.
- Use randomization device to acquire the truthful response from respondents.
- To estimate the Proportion of the population possessing sensitive characteristic.

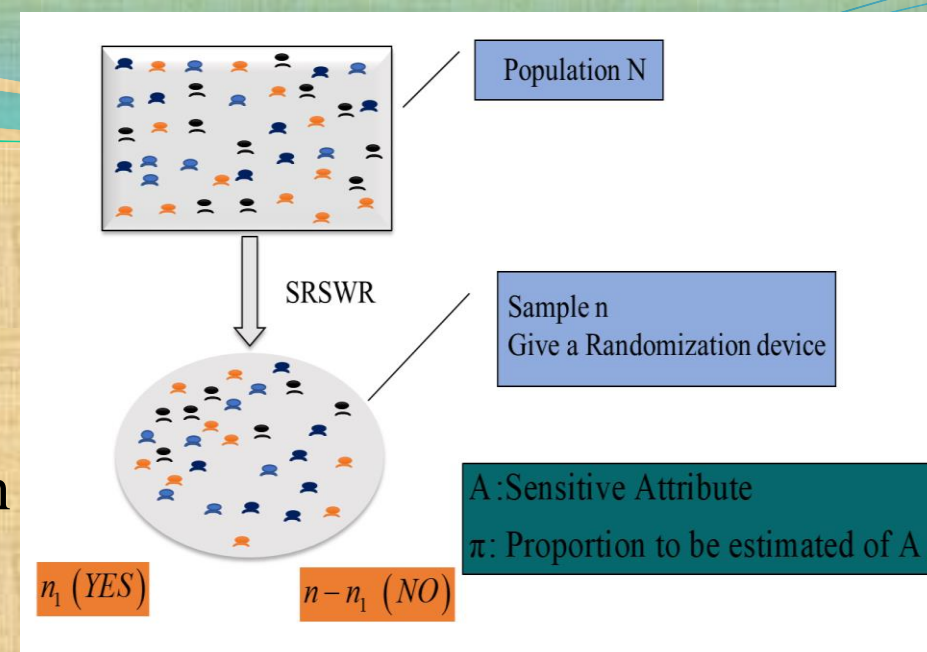
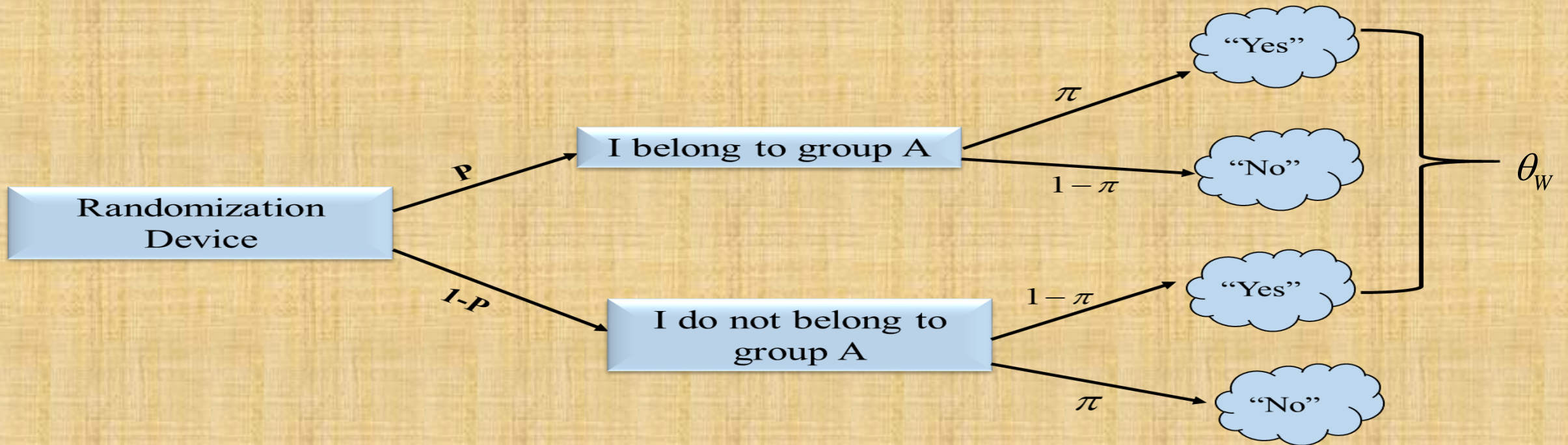




## ❖ Gradual Development: RRT

### ❖ Warner (1965) Technique

- Each respondent is provided an identical randomization device as:





- With the help of a randomized device, the respondent replies only “Yes” or “No” answers in a random sample of n respondents. The probability of “Yes” answer:

$$\theta_w = P\pi + (1-P)(1-\pi)$$

- The unbiased estimator of population proportion

$$\hat{\pi}_w = \frac{\hat{\theta}_w - (1-P)}{2P-1} \quad P \neq 0.5$$

$$\hat{\theta}_w = \frac{n_1}{n}$$

where  $\hat{\theta}_w$  is the observed proportion of "Yes" answer in the sample of n units drawn by the SRSWR sampling.

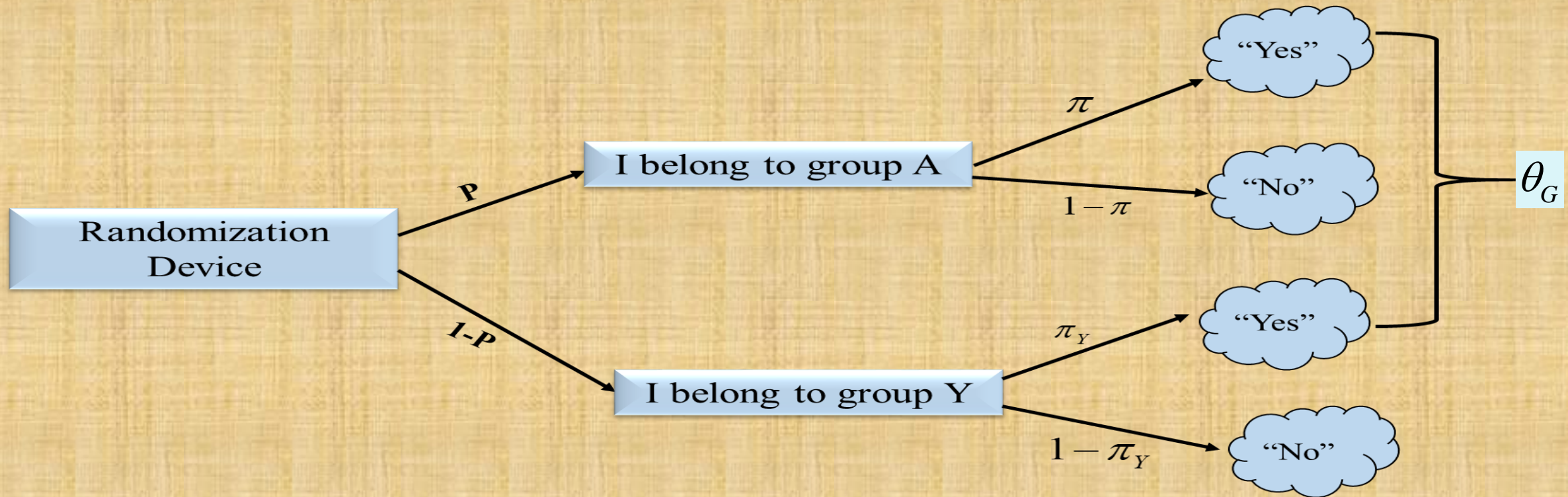
- The Variance is :

$$V(\hat{\pi}_w) = \frac{\pi(1-\pi)}{n} + \frac{P(1-P)}{n(2P-1)^2}$$

## ❖ Greenberg et al. (1969) Technique



- Each respondent is provided an identical randomized device as:



- For example, in estimating the proportion of persons having the extra marital relations in a certain community the two questions may be:

- i. Are you having extra marital relations?
- ii. Did you born in the month of March?

### ❖ **When the Proportion of unrelated character is known**

- With the help of a randomized device, the respondent replies only “Yes” or “No” answers in a random sample of n respondents. The probability of “Yes” answer in the population:

$$\theta_G = P\pi + (1 - P)\pi_Y$$



Let  $n_1$  be the number of observed "Yes" answer in the sample of  $n$  units. So that  $\hat{\theta}_G = \frac{n_1}{n}$

- The unbiased estimator of  $\pi$  is:

$$\hat{\pi}_G = \frac{\hat{\theta}_G - (1 - P)\pi_Y}{P}$$

- The variance of the estimator  $\hat{\pi}_G$  is:

$$V(\hat{\pi}_G) = \frac{\theta_G - (1 - \theta_G)}{nP^2}$$

Population: ( $N < \infty$ ); A: Sensitive Attribute

Y: Non-Sensitive Unrelated Attribute

$\pi$ : Proportion to be estimated of A

$\pi_Y$ : Proportion of Non-Sensitive unrelated attribute Y

## ❖ When the Proportion of unrelated character is unknown

Here  $\pi_Y$  the proportion of unrelated character Y in the population is unknown

- In this case the probability of “Yes” answer:

Probability of "yes" answer using device 1

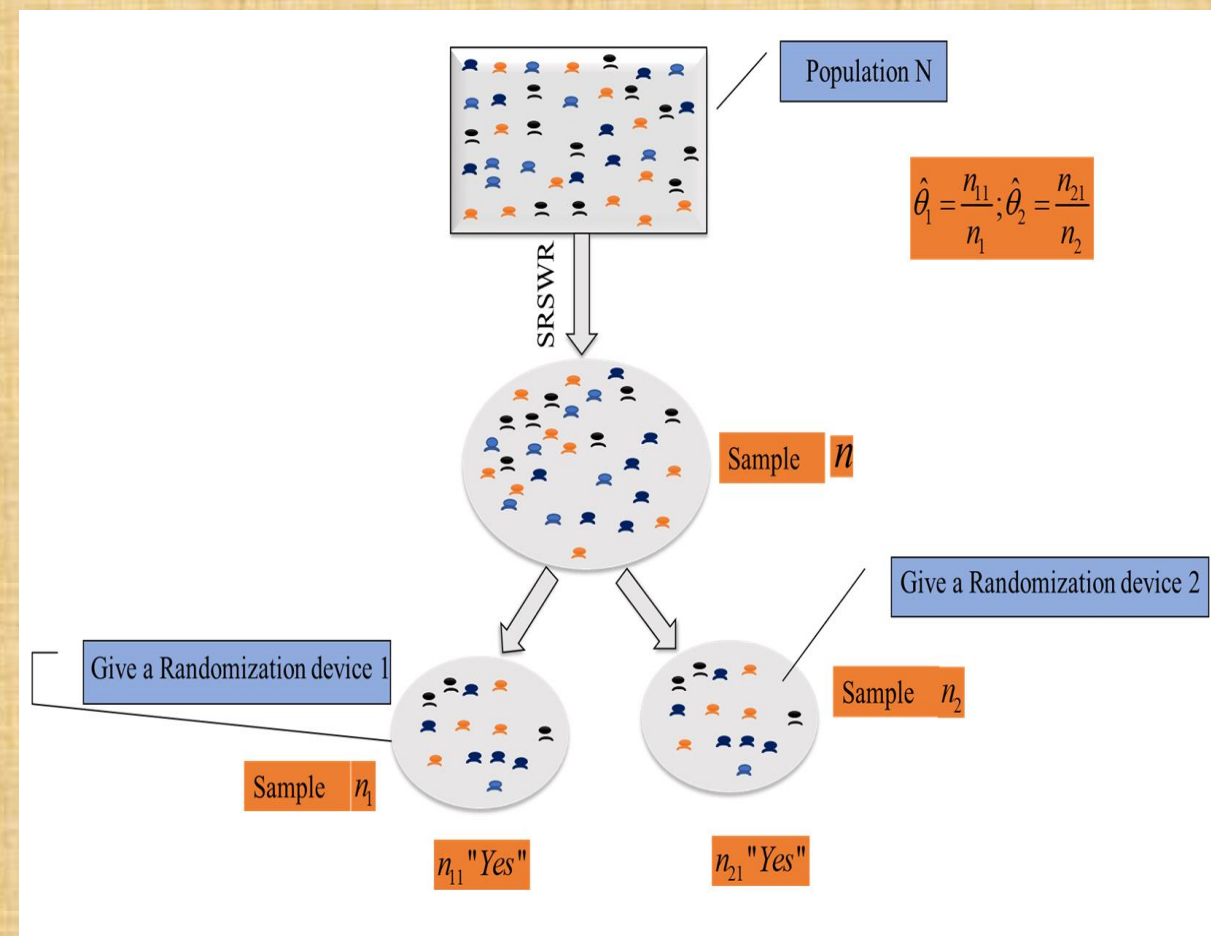
$$\theta_1 = P_1\pi + (1 - P_1)\pi_Y$$

Probability of "yes" answer using device 2

$$\theta_2 = P_2\pi + (1 - P_2)\pi_Y$$

- The unbiased estimator of  $\pi$  is:

$$\hat{\pi}_G = \frac{(1 - P_2)\hat{\theta}_1 - (1 - P_1)\hat{\theta}_2}{P_1 - P_2}$$



where  $\hat{\theta}_1$  and  $\hat{\theta}_2$  are the observed proportion of "Yes" answer in the first and second sample respectively.

- The variance of the estimator  $\hat{\pi}_G$  is:

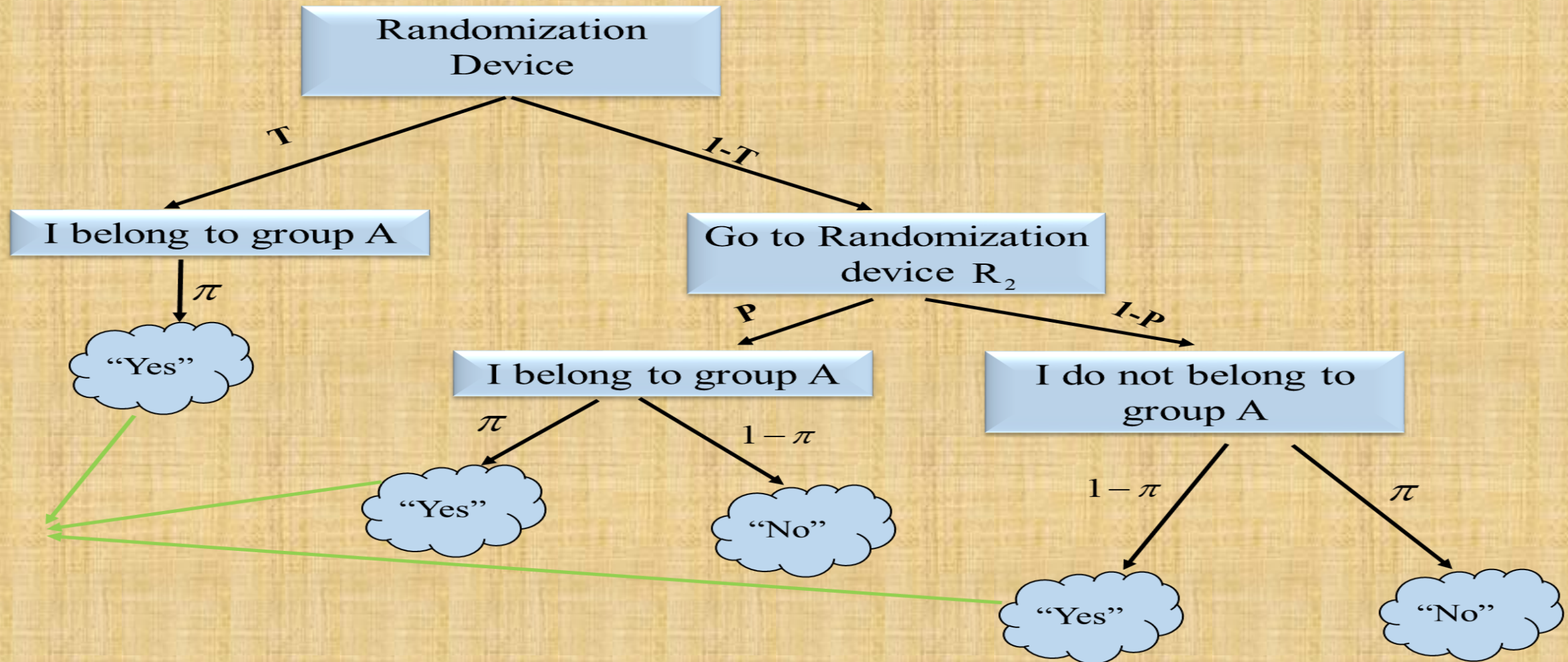
$$V(\hat{\pi}_G) = \frac{1}{(P_1 - P_2)^2} \left[ \frac{(1 - P_2)^2 \theta_1 (1 - \theta_1)}{n_1} + \frac{(1 - P_1)^2 \theta_2 (1 - \theta_2)}{n_2} \right]$$

For the best choice of  $n_1$  and  $n_2$ , it should follow the relation

$$\frac{n_1}{n_2} = \sqrt{\frac{\theta_1 (1 - \theta_1)}{\theta_2 (1 - \theta_2)} \frac{(1 - P_2)}{(1 - P_1)}}$$

## ❖ Mangat and Singh (1990) Technique

- Each respondent is provided an identical randomized device





- With the help of a randomized device, the respondent replies only “Yes” or “No” answers in a random sample of  $n$  respondents. The probability of “Yes” answer:

$$\theta_{MS} = T\pi + (1-T)\{P\pi + (1-P)(1-\pi)\}$$

$$\hat{\theta}_{MS} = \frac{n_1}{n}$$

- The unbiased estimator of  $\pi$  is:

$$\hat{\pi}_{MS} = \frac{\hat{\theta}_{MS} - (1-T)(1-P)}{2P - 1 + 2T(1-P)}$$

- The Variance is :

$$V(\hat{\pi}_{MS}) = \frac{\pi(1-\pi)}{n} + \frac{(1-T)(1-P)\{1-(1-T)(1-P)\}}{n\{2P-1+2T(1-P)\}^2}$$

Population:  $(N < \infty)$

A: Sensitive Attribute

$\pi$ : Proportion to be estimated of A



THANK  
YOU