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# 36-303: Sampling, Surveys and Society

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Quality in Surveys  
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# Handouts

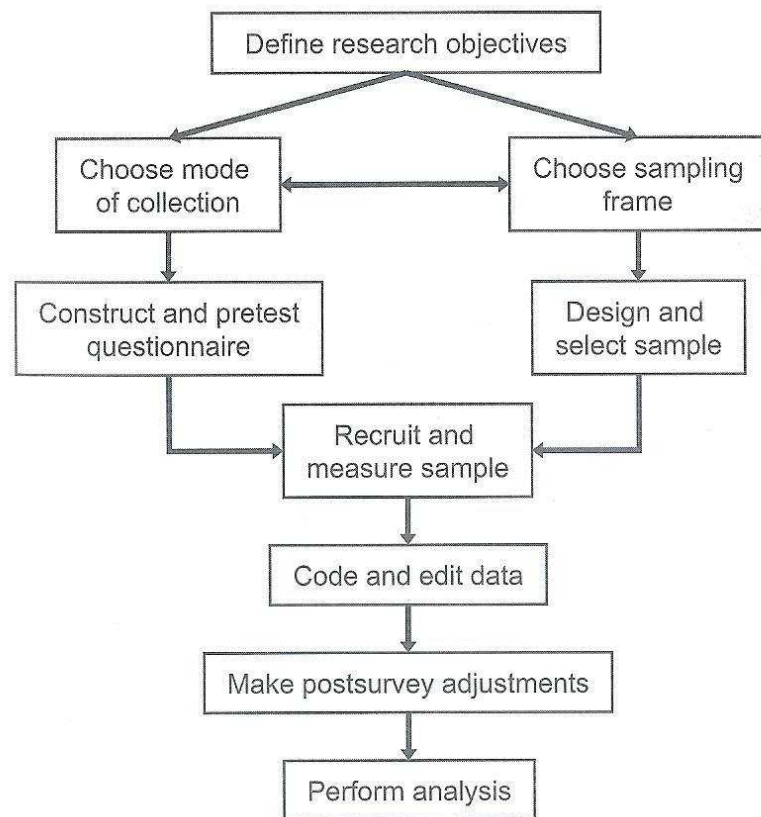
- Lecture Notes
- Examples of I.1 Proposals  
[due Tue Jan 25]

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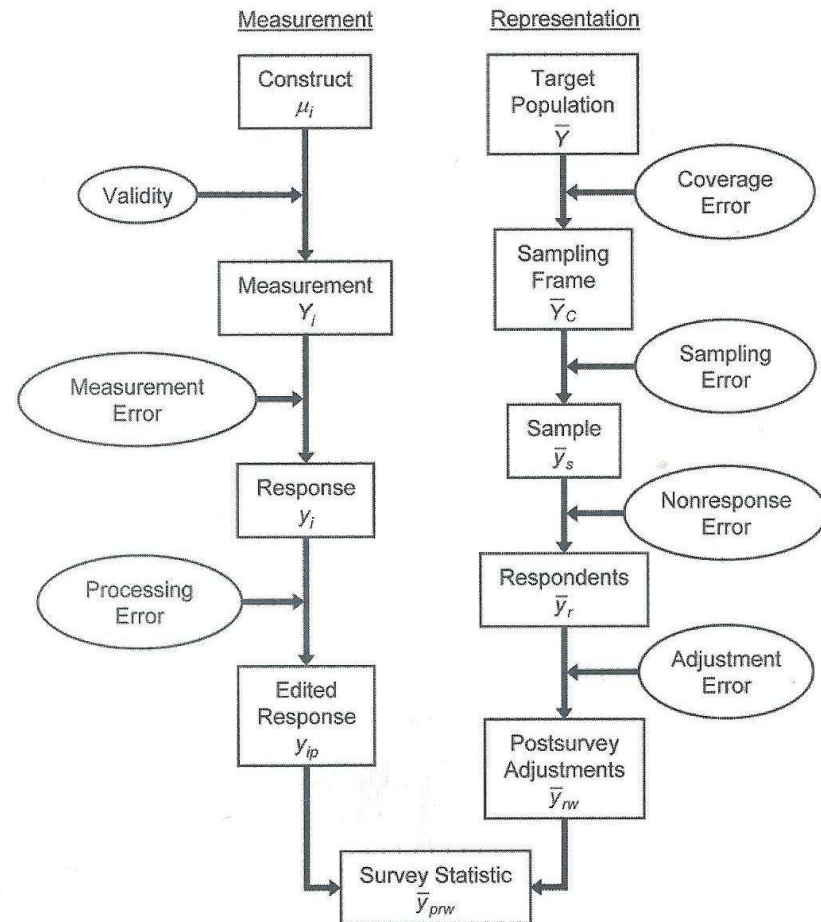
# Outline

- Quality in Surveys
- Project Proposals: I.1 on the “Project Schedule” handout.
- Reading:
  - Up to today: responsible for Groves Ch’s 1, 2, 3
  - Next week:
    - Groves Ch 5
    - Groves, Ch 11 (sections 1-6)
    - Groves Ch 4 (sections 1-3; we will do more later)
- Lecture notes online at  
<http://www.stat.cmu.edu/~brian/303>

# Quality in Surveys

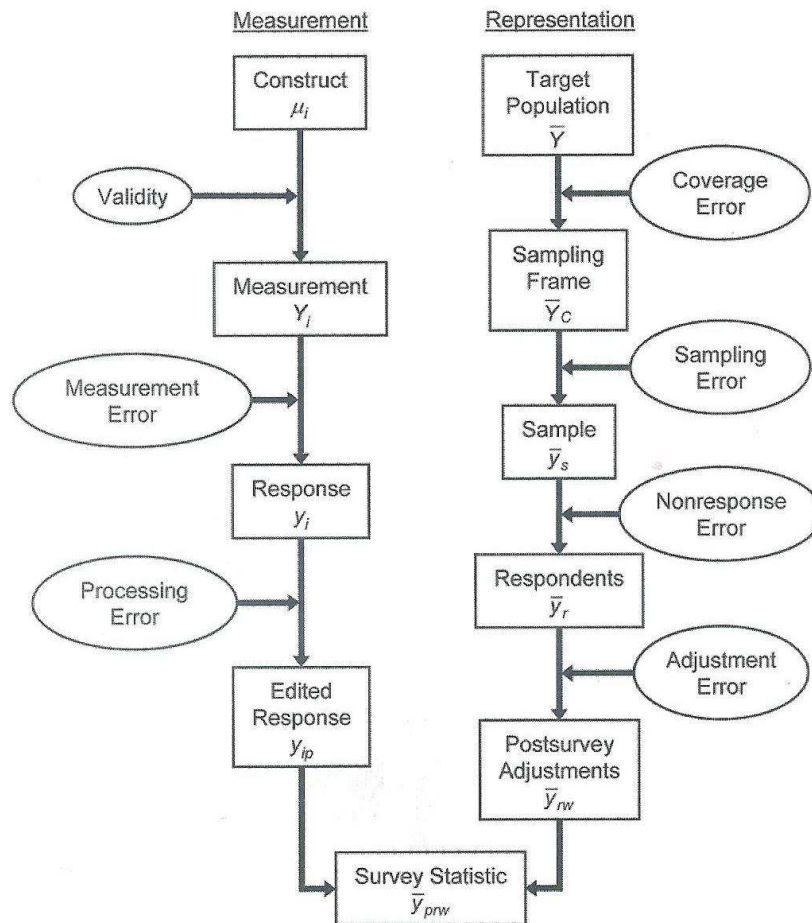


Process Perspective on Surveys



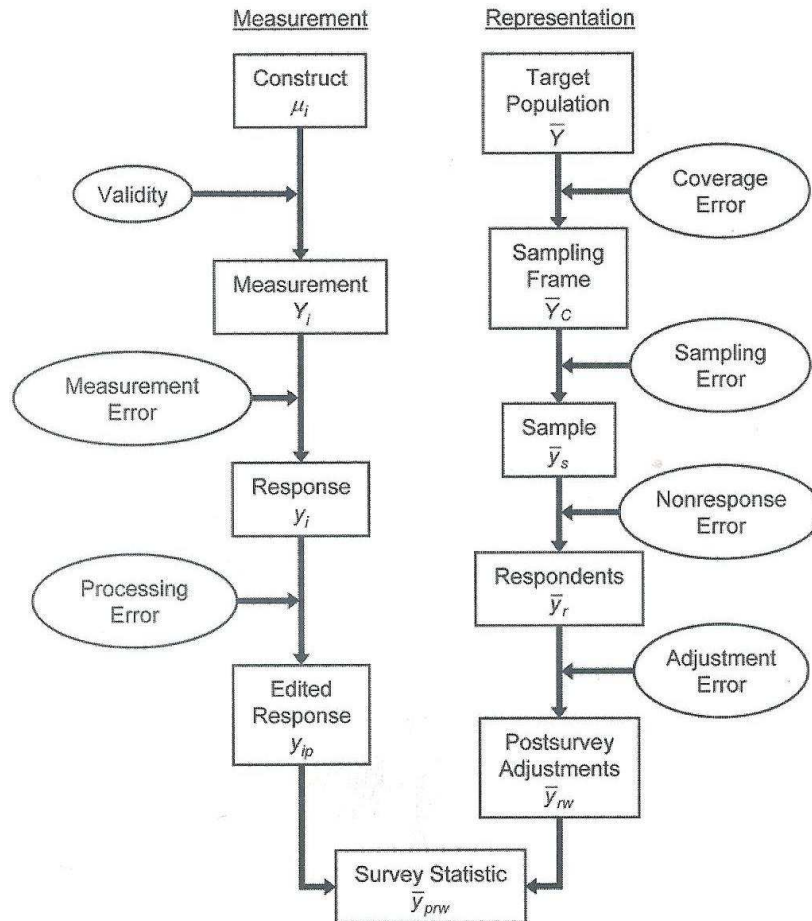
Quality Perspective on Surveys

# Quality Overview



- **Total Survey Error**
  - Each of the **Quality Components** has a verbal description and a statistical formulation
  - The **Quality Components** are properties of individual survey design and analysis decisions, not of whole surveys
- Our job is to make decisions to minimize error / maximize quality

# Measurement Quality



- Working down the left side:
  - *Validity*
  - *Measurement Error*
  - *Processing Error*

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## Some Notation...

- $\mu_i$  = value of the construct. E.g. # of doctor visits for  $i^{\text{th}}$  person in population,  $i=1, \dots, N$
- $Y_i$  = ideal value of the measurement for the  $i^{\text{th}}$  person in the sample,  $i=1, \dots, n$
- $y_i$  = observed value (reported number of doctor visits) for  $i^{\text{th}}$  sample person
- $y_{ip}$  = observed value after editing/processing
- $y_{it}$  = value on the  $t^{\text{th}}$  “trial” ( $t^{\text{th}}$  time we run the survey)

# Validity

- $Y_i = \mu_i + \epsilon_i$ 
  - $\mu_i$  is the “true value” for the population
  - $Y_i$  is the “ideal measured” value
  - $\epsilon_i$  is how much  $Y_i$  “deviates” from  $\mu_i$
  - Deviation/error is natural. We just have to account for it
- If there are  $T$  trials (repeats of the survey),  $t=1, \dots, T$ , we might write

$$Y_{it} = \mu_i + \epsilon_{it}$$

And expect that the errors  $\epsilon_{it}$  would “average out” over trials...

- A measure of the size of the errors  $\epsilon_i$  is

$$\text{Corr}(Y_i, \mu_i)$$

This correlation is a measure of the **Validity** of the measurement



# Measurement Error

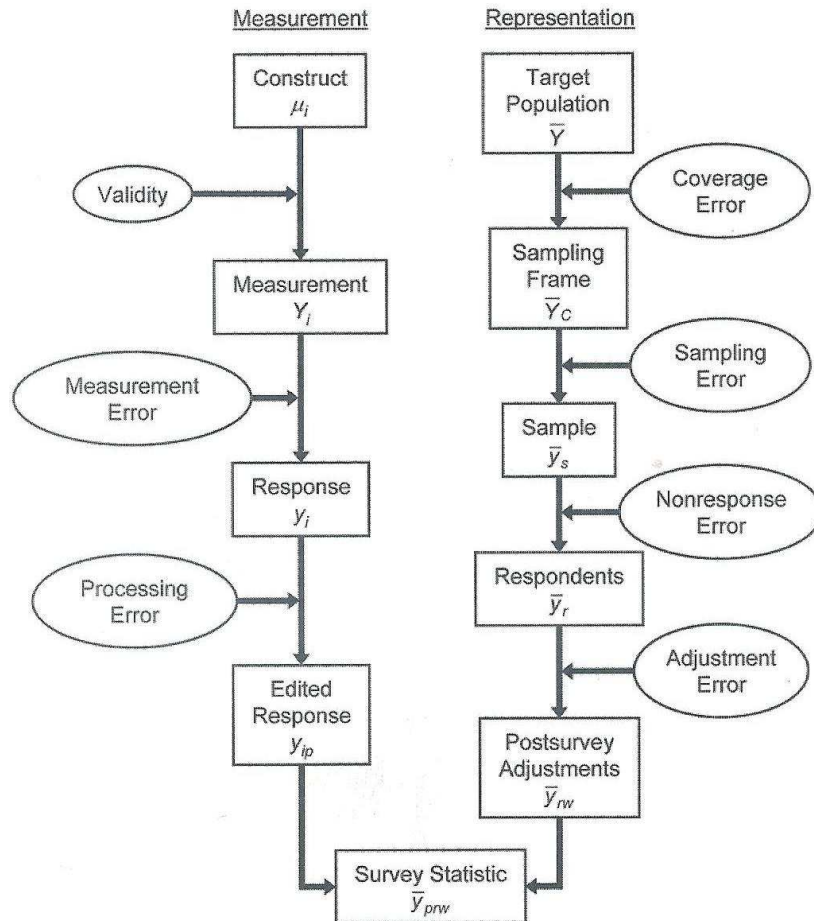
- $y_i - Y_i$  is the measurement error
  - $Y_i$  is the ideal measurement
  - $y_i$  is the observed measurement
- There are two kinds of measurement error to worry about
  - Variability:  $y_i = Y_i + \text{error}_i$ , and the error “averages out” over repeated trials:  $E_t[y_{it}] = Y_i$
  - Bias:  $y_i = Y_i + \text{something}$  that doesn’t “average out”:  $E_t[y_{it}] \neq Y_i$

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# Processing Error

- $y_{ip} - y_i$  is the processing error
  - $y_{ip}$  is the response after editing/processing
  - $y_i$  is the 'raw' response to the measurement
- These errors come in when you have to code, check, or fix survey responses, e.g.
  - Coding a verbal response
  - Range check – can this person have been in High School for 7 years?
  - Clumping, e.g. “income between \$10,000 and \$30,000”
- These are generally bias and not variability issues

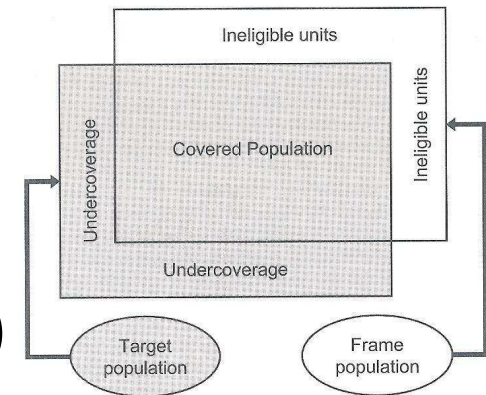
# Representation Quality



- Working down the right side:
  - ❑ Coverage Error
  - ❑ Sampling Error
  - ❑ Nonresponse Error (later lecture)
  - ❑ Adjustment Error

# Coverage Error

- $N$  = total Target Population (size)
- $C$  = target population covered in frame
- $U$  = target population missed by frame
- $\bar{Y}$  = mean of target population
- $\bar{Y}_C$  = mean of covered population
- $\bar{Y}_U$  = mean of uncovered population
- $\bar{Y}_C - \bar{Y} = \underline{\text{coverage error}}$ 
  - Also called Coverage Bias

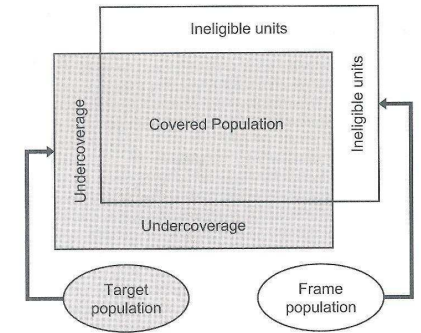


# Coverage Error (Cont'd)

$$\boxed{\bar{Y}_C - \bar{Y} = \frac{U}{N}(\bar{Y}_C - \bar{Y}_U)}$$

$$\bar{Y} = \frac{1}{N} \sum_{i=1}^N Y_i = \frac{1}{N} \left( \sum_C Y_i^C + \sum_U Y_i^U \right)$$

$$\begin{aligned} \bar{Y}_C - \bar{Y} &= \frac{1}{C} \sum_C Y_i^C - \frac{1}{N} \sum_{i=1}^N Y_i \\ &= \frac{1}{C} \sum_C Y_i^C - \frac{1}{N} \left( \sum_C Y_i^C + \sum_U Y_i^U \right) \\ &= \left( \frac{1}{C} - \frac{1}{N} \right) \sum_C Y_i^C - \frac{1}{N} \sum_U Y_i^U \\ &= \frac{U}{NC} \sum_C Y_i^C - \frac{U}{N} \cdot \frac{1}{U} \sum_U Y_i^U \\ &= \frac{U}{N} (\bar{Y}_C - \bar{Y}_U) \end{aligned}$$



# Coverage Error/Coverage Bias

- Suppose we are interested in Monthly Mortgage Payment (\$0 if you rent)
  - Total population is all adults in (US/Pgh/...)
  - Data collection method is random digit dialling
  - Sampling frame is callable land-line phone #'s
- Renters may be more likely to have only a cell phone than homeowners
  - Renters are undercovered by our frame
  - Our estimate of mean mortgage payment will be too high
  - If we can get an estimate of  $\frac{U}{N}(\bar{Y}_C - \bar{Y}_U)$   
Then we can estimate  $\bar{Y}_C - \bar{Y}$  and fix the bias!

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# Sampling Error

- How well does the sample represent the sampling frame?
  - Sampling bias
    - Best to try to anticipate and avoid
    - Can be looked at similarly to coverage bias
    - Another way to deal with is with weights, but this can introduce “adjustment error” (more in a couple pages)
  - Sampling variability – this is a more familiar issue! (see next page)

# Sampling Variability

- $\bar{y}_s = \frac{1}{n_s} \sum_{i=1}^{n_s} y_{si}$  is the *mean of the sample*
- $\bar{Y}_C = \frac{1}{C} \sum_C Y_i^C$  is the *mean of the frame*

The *Standard Error* for estimating  $\bar{Y}_C$  with  $\bar{y}_s$  is

$$SE = \sqrt{\frac{1}{S} \sum_{s=1}^S (\bar{y}_s - \bar{Y}_C)^2}$$

in case of *simple random sampling* (next week!) we know that

$$SE = SD / \sqrt{n_s} = \frac{\sqrt{\frac{1}{n_s - 1} \sum_{i=1}^{n_s} (y_{si} - \bar{y}_s)^2}}{\sqrt{n_s}}$$



# Adjustment Error

- This usually comes in the forms of weights.
- If the proportion of units in the sample is systematically different from the population, we may weight each unit:

$$\bar{y}_w = \frac{\sum_{i=1}^{n_s} w_i y_i}{\sum_{i=1}^{n_s} w_i}$$

- The main issues are (again) bias and variability of this estimate  $\bar{y}_w - \bar{Y}$

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# More on the Project Outline Handout

- Some Examples of Proposals (I.1, due Jan 25)
- Shall we look at the whole project outline as well?
  - This is your chance to ask questions about any parts of the handout that you read, and are concerned about.

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# Review

- Quality in Surveys
- More on the Project Outline Handout
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# Review

