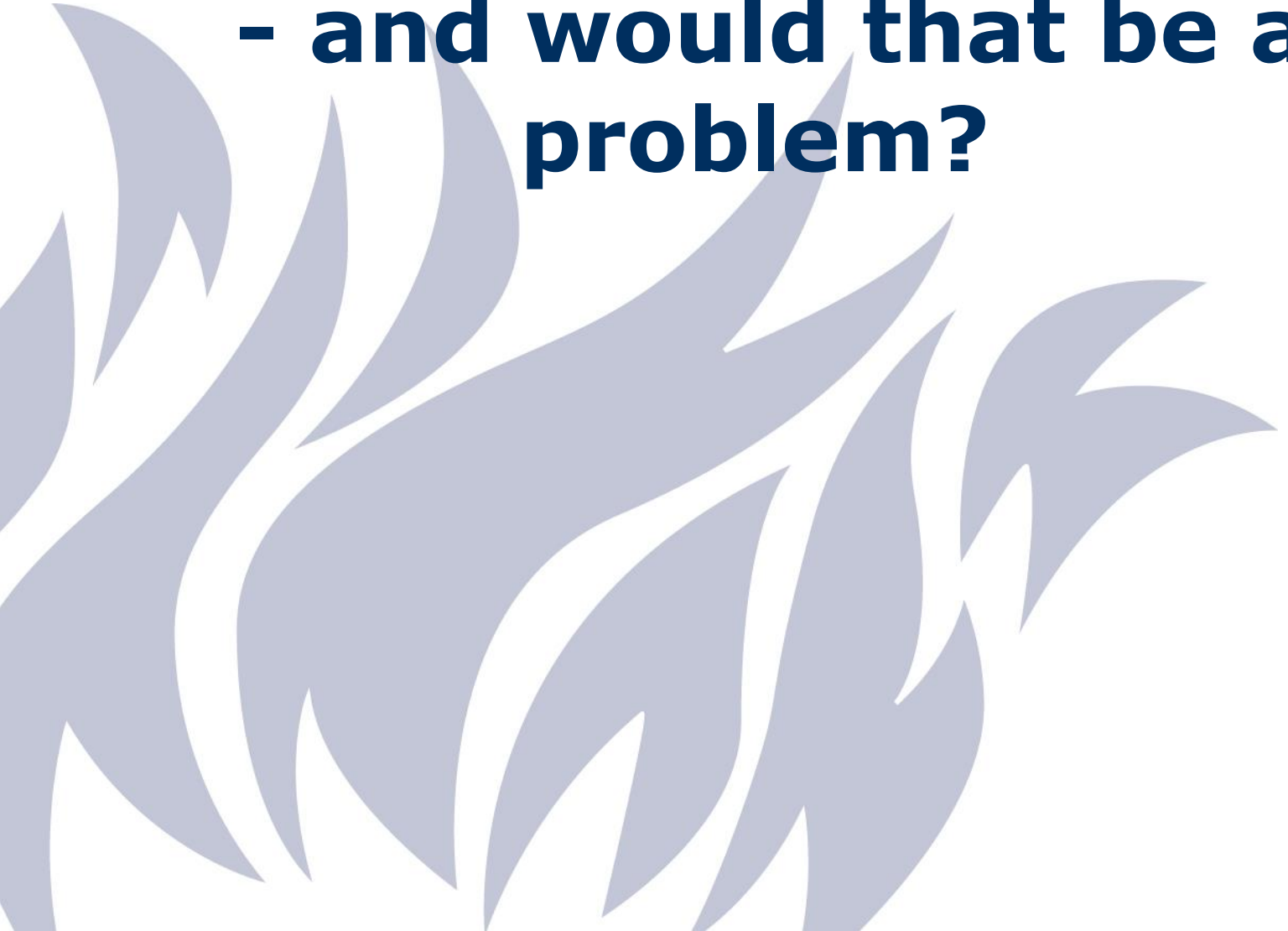


# Are we witnessing the end of random sampling in surveys?

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**- and would that be a  
problem?**



## Surveys have strong emphasis on representation/generality

(Kish 1987 page 7, Baker et al. 2013 sec 8.2, O'Muircheartaigh 1999)

# Reasons for random sampling

## Impartiality/transparency

- **Transparently impartial (cf referee tossing a coin at start of game)**
- **Easy to communicate**
- **As distant from human mind as it can be (random selection as the most stupid selection method conceivable)**
- **A judgement sample opens the Pandora's box of "improvements" of the realised sample**
- **Easy to accept "bad luck" in a random sample. In a judgement sample, there is no such a thing as 'bad luck'.**

## **Further reasons: Statistical culture / Knowledgebase**

- **Coordinated samples, PRNs**
- **Incorporated in university courses**
- **Lots of literature, including textbooks at various levels**
- **Lots of experience**
- **Knowledge among practitioners**
- **Public acceptance**

# Disadvantages with random sampling

- **Persuading reluctant respondents in a longitudinal survey may create attrition**
- **Or may create nonresponse in other surveys**  
(Bergman and Brage 2008)
- **Is it cost-effective? Strong emphasis on representation/generality means that expensive units need to be sampled and approached.**
- **Long period of field work**
- **Increases general response burden**

## Now focus on three sampling methods

1. Random sample from a perfect frame
2. Non-random sample from a perfect frame
3. Non-random sample from a frame that suffers from missing objects
  - Nonresponse in all three

# Inference in surveys

- **Sample  $s$**
- **Specify  $f(Y|X; \beta)$  for all variables  $Y$  for all  $N$  units.  $X$  to be used for sampling design and estimation.**
  1. **Analytic aim: inference about  $\beta$**
  2. **Descriptive aim: inference about  $Y_{\bar{s}}$  (complement)**
- **We need  $f(Y|X; \beta)$  for inference about  $Y_{\bar{s}}$**
- **Further, for 3,  $f(Y|X, Z; \beta)$ ,  $Z$  indicates availability on the frame (binary: missing/not missing)**



**1. Sample selection ignorability criterion:**

$$f(\mathbf{I}_S | \mathbf{Y}_S, \mathbf{X}) = f(\mathbf{I}_S | \mathbf{X})$$

**• True also for many nonrandom sampling designs**

(Little 1982, Smith 1983)

**2. To be able to ignore nonresponse:**

$$f(J_r | \mathbf{I}_S, \mathbf{Y}_S, \mathbf{X}) = f(J_r | \mathbf{I}_S, \mathbf{X})$$

**3. To be able to ignore frame availability:**

$$f(\mathbf{Y} | \mathbf{Z}, \mathbf{X}; \boldsymbol{\beta}) = f(\mathbf{Y} | \mathbf{X}; \boldsymbol{\beta})$$

(Little 1982, Smith 1983, Valliant et al. 2003)

- **We restrict attention to ignorable sampling designs; nonrandom samples must be ignorable**
- **I will argue: it is sampling method 3 that is different.**
- **Does criterion 3 hold in practice? Sometimes it does, sometimes it does not. See e.g. Baker et al. (2013), Callegaro et al. (2014), Craig et al. (2013), Gotway Crawford (2013), Erens et al. (2014), Martinsson (2013), Sjöström (2012), Yeager et al. (2011), etc.**

# On balanced samples

- What is balance?
- $\bar{x}_s = \bar{x}_r$  “Response set balance”

# A balanced sample is good to have

- $\bar{y}_s - \bar{y}_r = (\bar{x}_s - \bar{x}_r)' \hat{\beta}_r + (\hat{\beta}_s - \hat{\beta}_r)' \bar{x}_s$   
(Särndal & Lundquist 2014)
- **Does small  $(\bar{x}_s - \bar{x}_r)$  imply small  $(\hat{\beta}_s - \hat{\beta}_r)$ ?**
- **The answer is “in most cases, probably yes”**  
(Särndal & Lundquist 2014, Sec. 6)

- **So what is the causal mechanism**  
small  $(\bar{x}_s - \bar{x}_r)$   $\rightarrow$  small  $(\hat{\beta}_s - \hat{\beta}_r)$ ?

**Loosely speaking, it is: Small variance of response propensities (in groups defined by  $x$ )** (Särndal & Lundquist 2014)

- **Note that we know  $\bar{x}_s, \bar{x}_r$  and that we can manipulate  $\bar{x}_r$  by adaptive sampling**  
(Schouten et al. 2013, Särndal & Lundquist 2014)
- **They start with a random sample. Necessary?**

# Paralleling the reasoning in Särndal & Lundquist (2014)

Let

- $U$  be the population, with  $x$  known for all units
- $r$  be a non-random sample
- $s$  a random sample *that you never drew*

Define

- $t = (\bar{x}_r - \bar{x}_U)' b_r$  (known)
- $v = (b_r - b_s)' \bar{x}_U$  (unknown; it is  $b_s$  that is unknown)

where...

- $J_k = \begin{cases} 1 & \text{if } k \in r \\ 0 & \text{if } k \notin r \end{cases}$
- $b_r = (\sum_S J_k \mathbf{x}_k \mathbf{x}'_k)^{-1} (\sum_S J_k \mathbf{x}_k y_k)$
- $b_s = (\sum_S \mathbf{x}_k \mathbf{x}'_k)^{-1} (\sum_S \mathbf{x}_k y_k)$
- **Aim: reduce**  $\bar{y}_s - \bar{y}_r = t + v$

- Often feasible (?) to reduce  $t$  by reducing  $\bar{x}_r - \bar{x}_U$  through some form of adaptive sampling
- If the  $J_k$  are constant, then  $b_r = b_s$
- If the variance among the  $J_k$  is small,  $b_r$  should be fairly close to  $b_s$
- Either reduce  $\bar{x}_r - \bar{x}_U$  or make the variance among the  $J_k$  small.



# Representativeness

- Meaning of 'a response set is **representative** (of some target population)'?
- Roughly: zero variance among the  $J_k$  (Schouten et al. 2012)
- Barry Schouten and co-workers proposed an 'R-indicator' to follow variance among the  $J_k$  while collecting data

# Conclusions and further issues

- **Suppose you are successful in balancing the set of responses. Does it matter whether you have started from a random sample or a nonrandom, ignorable sample? It would seem that it does not.**
- **A more practical issue: If you strive for balancing the response set, is it easier to start from a random sample?**
- **What is best, balancing response set or adjusting through estimation? Some evidence that balancing is slightly better** (Schouten et al. 2014) **You can do both** (Särndal 2011)

- **Is it practicably doable to achieve balance?**
- **Of course, there is a broader picture** (Schouten et al. 2012)

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