

Surveys

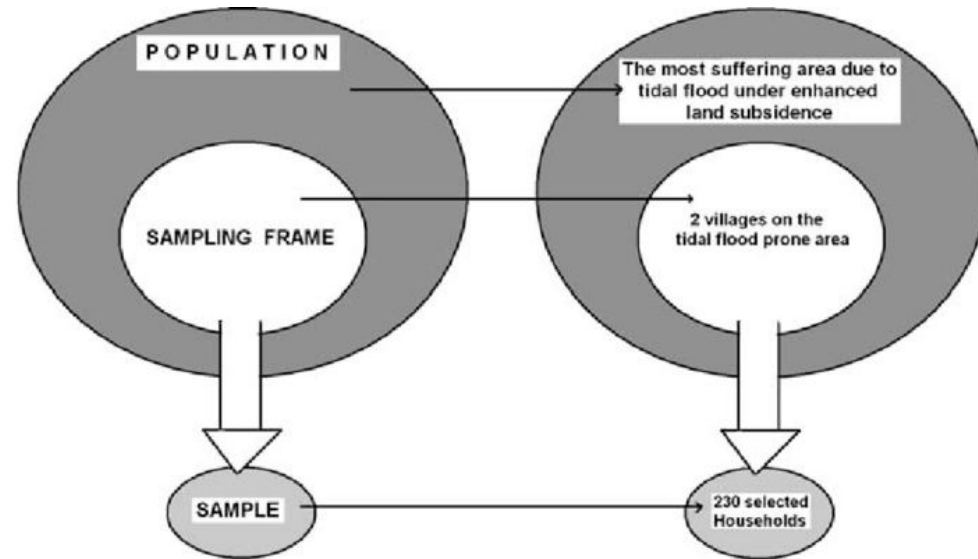
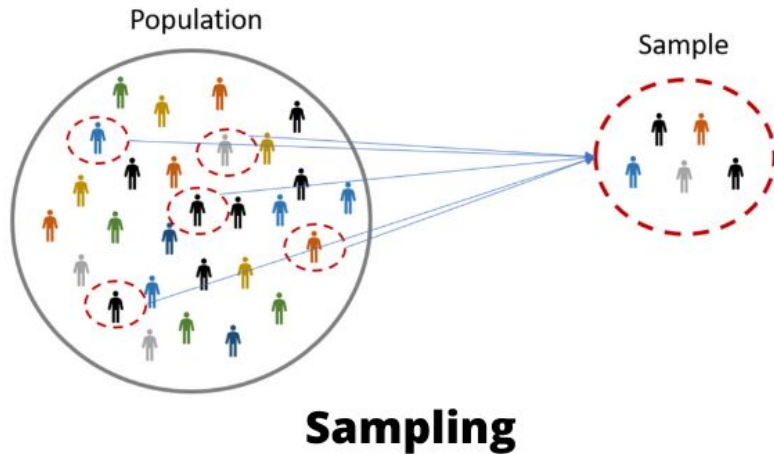
Asking Question

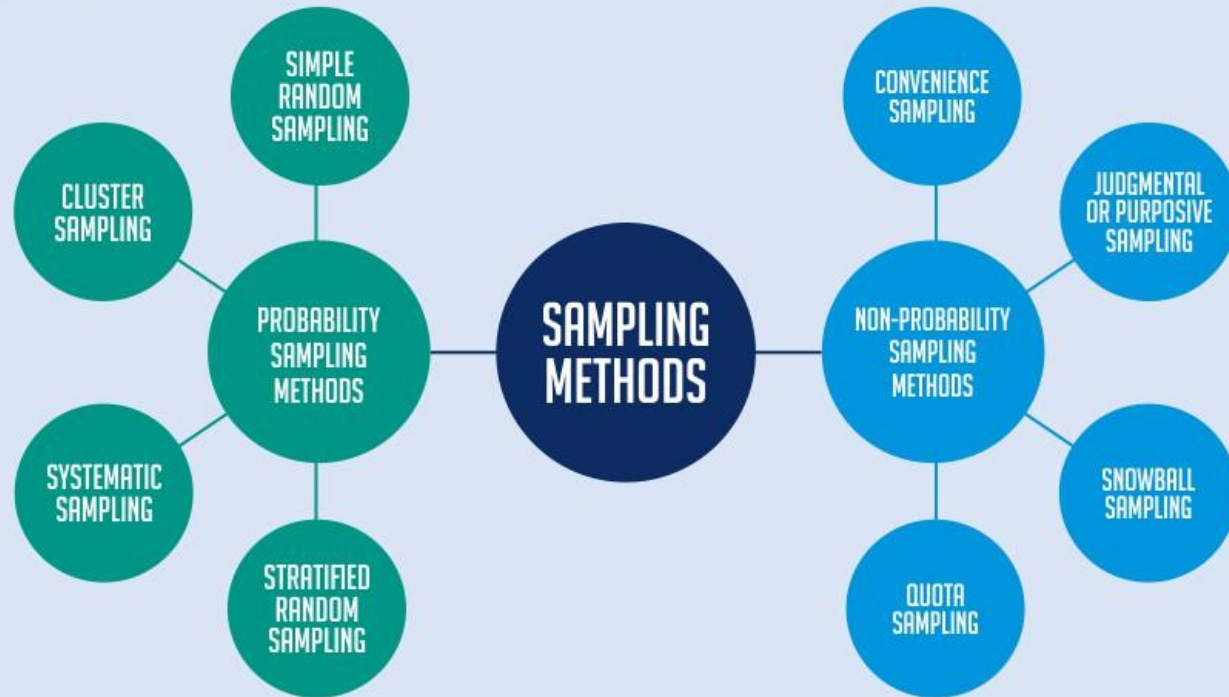
Computational Sociology
25 August 2021

History of survey

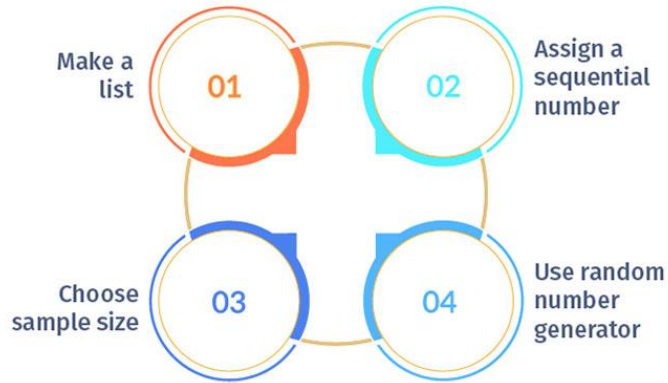
- What is sample survey?
- What is the benefit of survey?
- How to conduct survey?
- Three era of sample survey (Groves 2011)
 - 1930s-ff
 - 1970s-tel-RDD
 - Digital era

Population, sample and sample frame



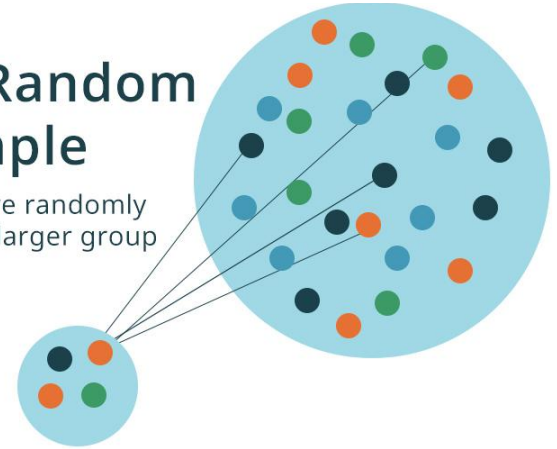


STEPS TO CONDUCT SIMPLE RANDOM SAMPLING



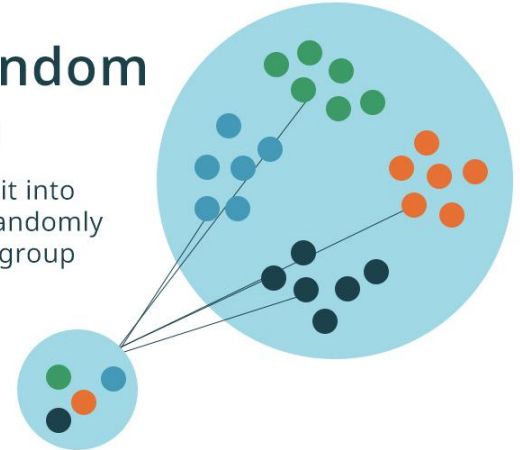
Simple Random Sample

Respondents are randomly selected from a larger group

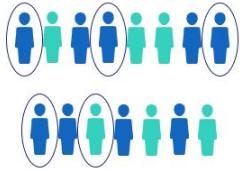


Stratified Random Sample

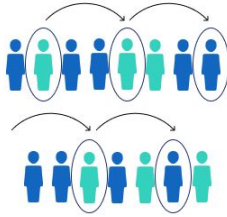
Respondents are split into sub-groups and then randomly selected from each group



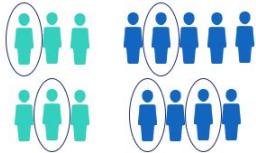
Simple random sample



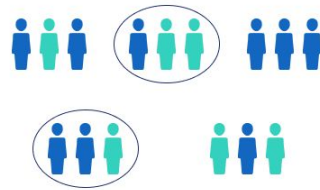
Systematic sample



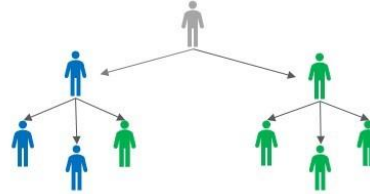
Stratified sample



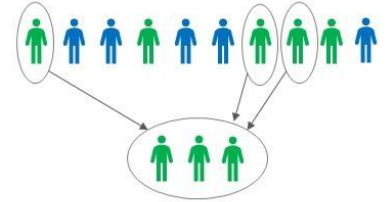
Cluster sample



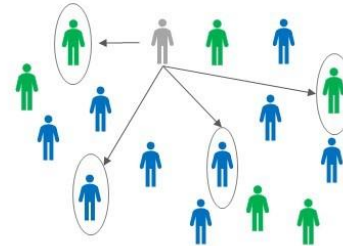
Snowball sample



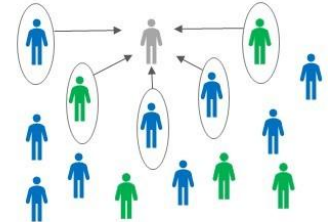
Quota sample



Judgement sample



Convenience sample



Three eras of survey research

	Sampling	Interviewing	Data environment
First era	Area probability sampling	Face-to-face	Stand-alone surveys
Second era	Random-digit-dialing (RDD) probability sampling	Telephone	Stand-alone surveys
Third era	Non-probability sampling	Computer-administered	Surveys linked to big data sources

Asking vs observing

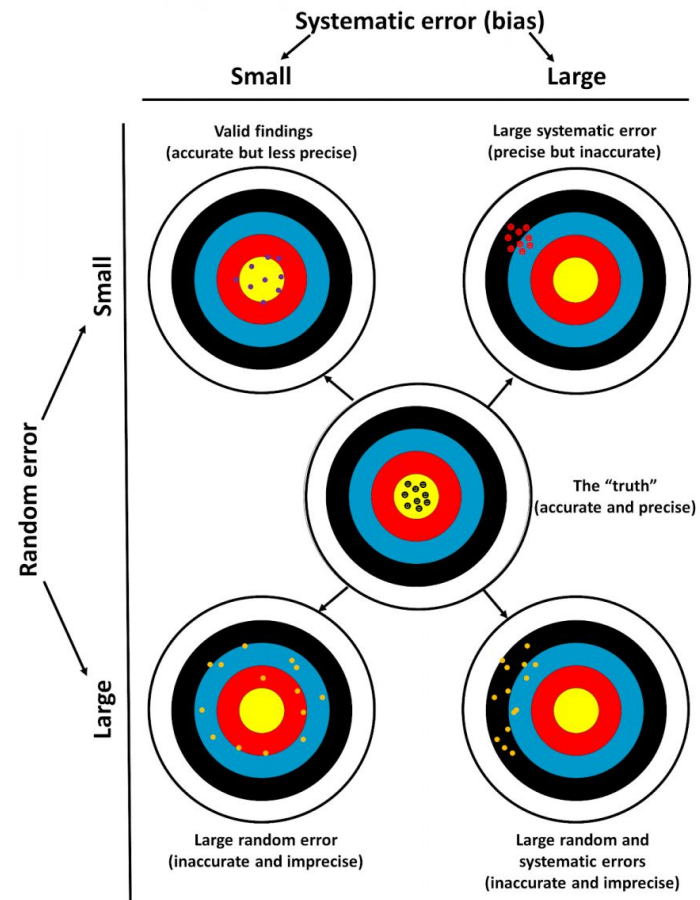
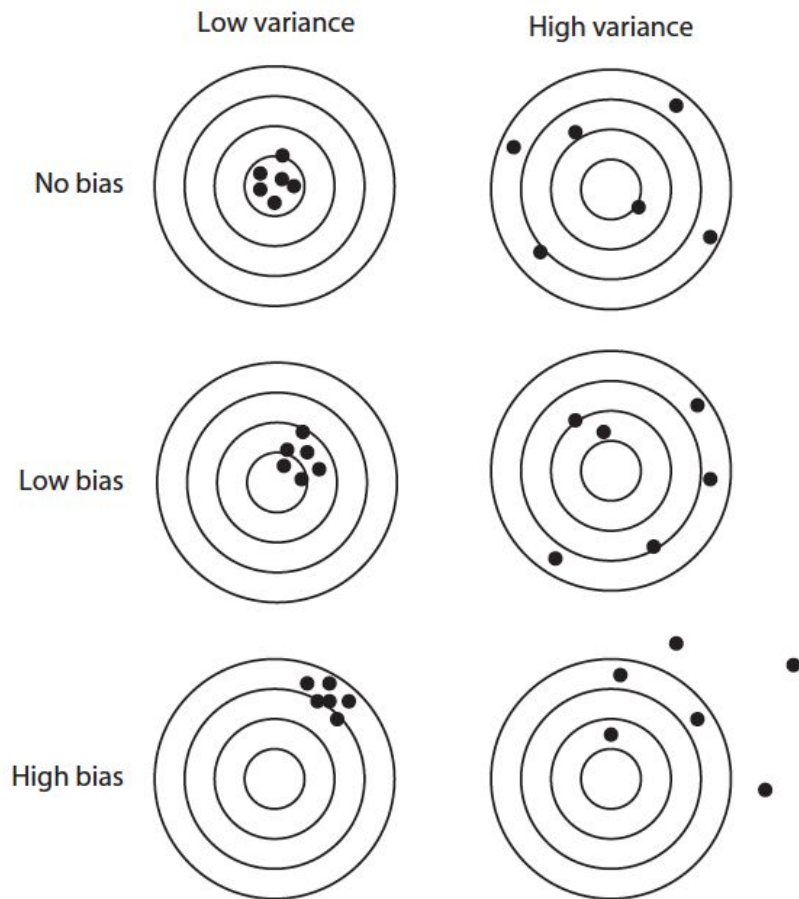
- What to ask?
- Participant observation
- Non-participant observation

The total survey error framework

Total survey error = representation errors + measurement errors

Bias and Error

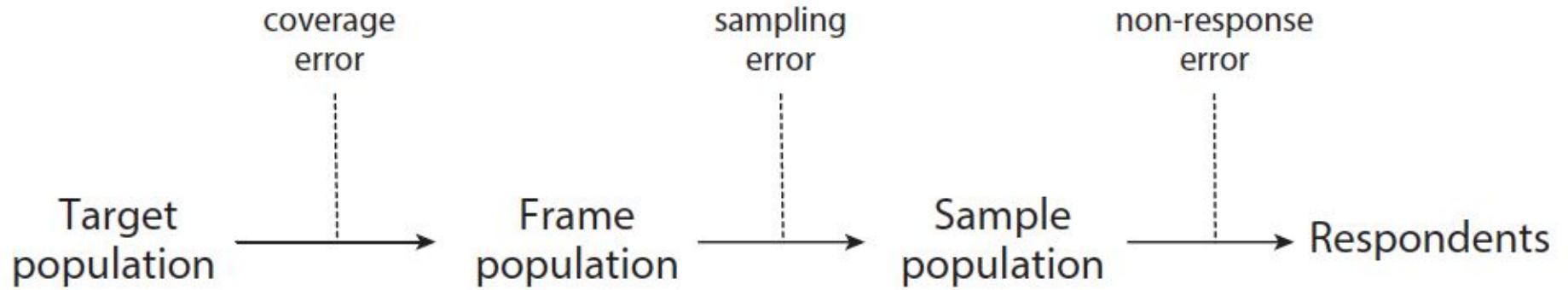
- Difference between error and bias: true error is random while bias is systematic error.
- Random sampling error can occur while selecting a population to survey due to uncontrollable issues.
- A probability survey must include a margin of error and a confidence level. This tells us how much effect random sampling error could have on a study's results.
- Bias cannot be measured statistically and it can give a skewed artificial result.



Representation

Representation is about making inferences from your respondents to your target population.

Representation error





Fighting poverty with data

Joshua Evan Blumenstock

+ See all authors and affiliations

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Policy-makers in the world's poorest countries are often forced to make decisions based on limited data. Consider Angola, which recently conducted its first postcolonial census. In the 44 years that elapsed between the prior census and the recent one, the country's population grew from 5.6 million to 24.3 million, and the country experienced a protracted civil war that displaced millions of citizens. In situations where reliable survey data are missing or out of date, a novel line of research offers promising alternatives. On page [790](#) of this issue, Jean *et al.* (1) apply recent advances in machine learning to high-resolution satellite imagery to accurately measure regional poverty in Africa.

Traditionally, wealth and poverty are measured through surveys of household income and consumption (2). These data provide a critical input to the world's most prominent antipoverty programs, from basic cash transfer programs to multifaceted aid programs designed to target the extreme poor (3). However, nationally representative surveys cost tens to hundreds of millions of dollars to collect, and many developing countries go for decades without updating their estimates.



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RESEARCH ARTICLE

Combining satellite imagery and machine learning to predict poverty

Predicting poverty

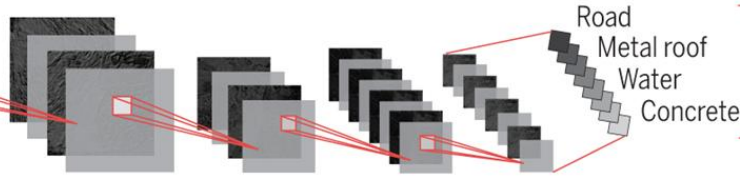
Satellite images can be used to estimate wealth in remote regions.

Neural network learns features in satellite images that correlate with economic activity

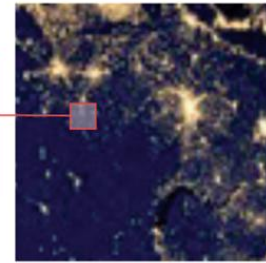
Daytime satellite photos capture details of the landscape



Convolutional Neural Network (CNN) associates features from daytime photos with nightlight intensity



Satellite nightlights are a proxy for economic activity

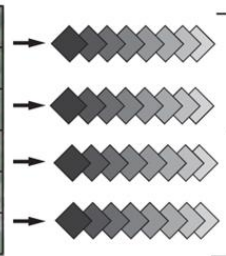
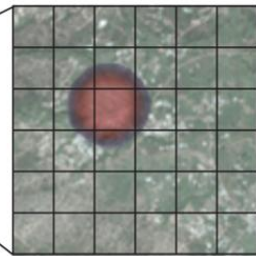


Daytime satellite images can be used to predict regional wealth

Household survey locations



CNN processes satellite photos of each survey site



Features from multiple photos are averaged



Ridge regression model reconstructs ground truth estimates of poverty