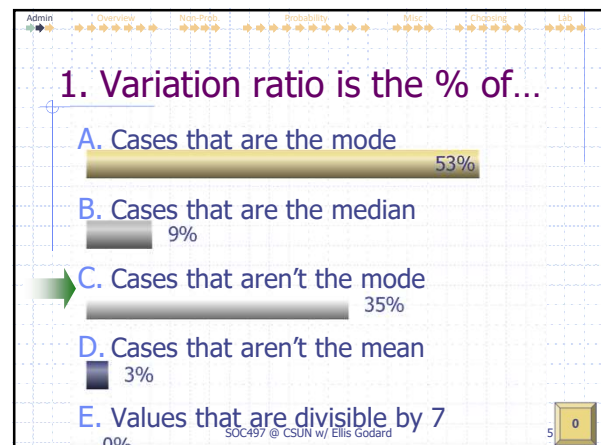


I do love me a schedule...

DEADLINES			LECTURE		LABS	
Date	Things Due	Reading (USD)	Category	TOPIC	Lab #	Lab Assignment
Mon Aug 26			Orientation	Orientation		Project Charter
Wed Aug 28			Orientation	History & Research		Research Design
Fri Aug 30			Orientation	Research Design		Research Design
Mon Sep 2			Orientation	SPSS & Intro		Computing (intro)
Wed Sep 4			Orientation	Researching Research (part)		Writing Paper
Fri Sep 6			Orientation	Researching Research (part)		Recording (colleges)
Mon Sep 9			Orientation	Researching Research (part)		Recording (colleges)
Wed Sep 11			Orientation	Researching Research (part)		Recording (colleges)
Fri Sep 13			Orientation	Researching Research (part)		Recording (colleges)
Mon Sep 16			Orientation	Researching Research (part)		Recording (colleges)
Wed Sep 18			Orientation	Researching Research (part)		Recording (colleges)
Fri Sep 20			Orientation	Researching Research (part)		Recording (colleges)
Mon Sep 23			Orientation	Researching Research (part)		Recording (colleges)
Wed Sep 25			Orientation	Researching Research (part)		Recording (colleges)
Fri Sep 27			Orientation	Researching Research (part)		Recording (colleges)
Mon Sep 30			Orientation	Researching Research (part)		Recording (colleges)
Wed Oct 2			Orientation	Researching Research (part)		Recording (colleges)
Fri Oct 4			Orientation	Researching Research (part)		Recording (colleges)
Mon Oct 7			Orientation	Researching Research (part)		Recording (colleges)
Wed Oct 9			Orientation	Researching Research (part)		Recording (colleges)
Fri Oct 11			Orientation	Researching Research (part)		Recording (colleges)
Mon Oct 14			Orientation	Researching Research (part)		Recording (colleges)
Wed Oct 16			Orientation	Researching Research (part)		Recording (colleges)
Fri Oct 18			Orientation	Researching Research (part)		Recording (colleges)
Mon Oct 21			Orientation	Researching Research (part)		Recording (colleges)
Wed Oct 23			Orientation	Researching Research (part)		Recording (colleges)
Fri Oct 25			Orientation	Researching Research (part)		Recording (colleges)
Mon Oct 28			Orientation	Researching Research (part)		Recording (colleges)
Wed Oct 30			Orientation	Researching Research (part)		Recording (colleges)
Fri Oct 31			Orientation	Researching Research (part)		Recording (colleges)
Mon Nov 3			Orientation	Researching Research (part)		Recording (colleges)
Wed Nov 5			Orientation	Researching Research (part)		Recording (colleges)
Fri Nov 7			Orientation	Researching Research (part)		Recording (colleges)
Mon Nov 10			Orientation	Researching Research (part)		Recording (colleges)
Wed Nov 12			Orientation	Researching Research (part)		Recording (colleges)
Fri Nov 14			Orientation	Researching Research (part)		Recording (colleges)
Mon Nov 17			Orientation	Researching Research (part)		Recording (colleges)
Wed Nov 19			Orientation	Researching Research (part)		Recording (colleges)
Fri Nov 21			Orientation	Researching Research (part)		Recording (colleges)
Mon Nov 24			Orientation	Researching Research (part)		Recording (colleges)
Wed Nov 26			Orientation	Researching Research (part)		Recording (colleges)
Fri Nov 28			Orientation	Researching Research (part)		Recording (colleges)
Mon Nov 30			Orientation	Researching Research (part)		Recording (colleges)
Wed Dec 3			Orientation	Researching Research (part)		Recording (colleges)
Fri Dec 5			Orientation	Researching Research (part)		Recording (colleges)
Mon Dec 7			Orientation	Researching Research (part)		Recording (colleges)
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SOC497/L: SOCIOLOGY RESEARCH METHODS

Sampling Effects:

Errors, Bias, & External Validity

Ellis Godard

Outline for Today

- ◆ Overview
 - Non-probability methods (6)
 - Probability methods (4)
- ◆ Census vs. Sample
- ◆ Choosing a Sample Size (time permitting)
- ◆ Lab: Sampling methods

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Admin Overview Non-Prob. Probability Misc. Chaining LSA

Why Sampling?

- ◆ Exercising a scientific vision
 - Want to order empirical facts
 - To do that, find abstractable patterns
 - ◆ Infer (generalize to) other cases from them
 - Do it iteratively, between data and theory
 - ◆ Wheel of science
- ◆ Need to sample - can't study all cases
 - Too expensive
 - Too time consuming
 - May not even know who the cases are

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Census Realities

- ◆ US census said to be "usually one of the least newsworthy of federal government activities"
 - 2000 Census, however, was front page news
 - Congress, Pres, Supreme Court, media, & others
- ◆ Several big issues, including
 - changing measurement of race (split Qs, *multiple response*, etc.)
 - whether to get rid of sampling
 - ◆ constitution requires an "actual enumeration" but undercounts and overcounts happen in practice
 - ◆ Census procedure = conduct TWO samples -> 2 datasets for all 7M census blocks
 - Long form / short form
 - weighting and adjustments
- ◆ Sampling (& apportionment) debates not new

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How Sampling Fits In

- ◆ Make inferences from sample to other cases
 - If follow rules, can use *sampling theory*
 - Allows inferences from observed to unobserved cases
- ◆ Next week starts Analysis & Presentation
 - What we DO w/ those facts
 - forms of analysis, esp. of quantitative data
 - particular focus on the elaboration method
- ◆ First: how we COLLECT those facts
 - inc. selecting cases and methods/forms of observation
 - start today w. the logic of sampling

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The Problem with Samples

- ◆ Remember sampling error:
 - Numeric difference btwn sample statistic & pop. parameter
 - Sample 50% male, population 48% male; *sampling error* = 2%
 - Avg educ level = 18 yrs in class, 11 in state; *sampling error* = 7
 - Warning: *not the same as standard error*
- ◆ *Want* the sample to look like the population
 - Problem:
 - ◆ usually don't *know* the parameter
 - Solutions:
 - ◆ *Estimate* parameters w/ ideas about errors across multiple samples
 - ◆ Ensure methodological rigor (inc. high response rate)
 - ◆ Appropriate sample size
 - Choices – esp the *kind* of sampling employed

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Census vs. Sample

- ◆ Sometimes can obtain data about entire pop.
 - census = when gather data on every member
 - not always possible (see Kish 1987:140)
- ◆ Advantages of censuses
 - precise, detailed to small domains, inclusive, credible
- ◆ Advantages of samples
 - rich, complex, diverse, flexible, accurate, pertinent, inexpensive
- ◆ U.S. census uses multi-stage clusters
 - phonebooks > pages > names;
 - or colleges > classes > students


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Non-Random Samples

Trump falsely claims every poll says he won the debate w Harris
<https://www.cnn.com/2024/09/12/politics/fact-check-trump-claims-won-debate>

He's citing a Newsmax poll (a terrible sample) AND he's misrepresenting *its* results



He wrote on social media on Wednesday afternoon: "Every Poll has us WINNING, in one case, 92-8, so why would I do a Rematch?"

"We won the debate according to every poll — every single poll, I think," Trump told reporters later in the afternoon.

Facts First: Trump's claim is false. As of Thursday morning, every major scientific poll about the debate — every poll that used random sampling techniques to try to obtain a representative picture of US public opinion — had found that Harris won. The polls in which most respondents said Trump won were unscientific junk polls — open questions posted online — that allowed an unlimited number of people to click and respond no matter how old they are or what country they live in, making the results useless as a measure of US public opinion.

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Two General Types of Samples

- 1. Probability**
 - probability of selection of each case is known
 - Not necessarily equal
 - More likely to be representative; no *systematic* bias
 - Probability theory estimates *degree* of representativeness
 - Few samples truly are, but we try as best as possible
- 2. Nonprobability**
 - Everything else; some bias or limitation
 - probability of selection *isn't* known
 - e.g. passersby on the street

◆ Not necessarily survey (e.g. fingerprints)

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Quota Sampling (nonprobability #6)

- ◆ Involves caps on certain proportions
- ◆ Start w/ desired target demographics
 - e.g. 50% F, 12% African-American
- ◆ The intersections are cell proportions
- ◆ The matrix of all those cells is a *quota frame*
- ◆ *Not* the same as strata or clusters (next)
 - Limits here are once the study begins
- ◆ Problems:
 - quota frame must be accurate
 - sampling errors are added for each *cell*
 - must be careful about biased selection w/i frame

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Non-probability Samples (3 of 6*)

- 1. Available Subjects**
 - Folks nearby – friends, colleagues, family
 - Adv/Disadv: Prone to bias, though okay for a pretest
- 2. Volunteer/Convenience**
 - Examples:
 - People at street corner, in library, at a mall, etc.;
 - comment cards, magazine inserts, etc.
 - Disadvantages: Biased to extremes (those w/ interest) & negatives (those w/ complaints)
 - Advantages: Oft used in marketing but not much scientific use
- 3. Purposive/Judgmental**
 - ID typical group based on knowledge of population & purpose of the study
 - Good for initial design w/ accessible subset
 - Good if focusing on deviant/trouble cases (what not fit gen. pattern)

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* Note: Only 4 of these are in Babble

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2. The opposite of probability sampling is...

- Conventional Sampling
0%
- Cluster Sampling
0%
- Modern Sampling
0%
- ☹️ D. Non-probability sampling
97%
- Statistical Sampling
3%

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Non-probability Samples

- 4. Informants**
 - May need them to be typical
 - But those willing to talk are not know what you need to know
 - may be powerful, public, or famous
- 5. Snowball**
 - Sample size accumulates (grows) through time (as it rolls on...)
 - start w/ known informants; ask for help in finding 2 other respondents
 - Esp. used for exploratory purposes
 - Also description, esp. when members are hard to locate or when sampling frame not complete
 - Homeless, AIDS patients in early 80s, immigrants, etc.

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* Note: Only 4 of these are in Babble



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Note on Representativeness

- ◆ Each of these methods introduces bias
 - Assumptions about availability of subjects
 - Assumptions about characteristics of pop.
- ◆ "Probability samples" asserts will get an *approximately* representative sample

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Probability Samples (4)

- ◆ Simple Random Sample
 - Random number table picks the cases
- ◆ Systematic Random Sample
 - Use a "skip number" to take every Kth case
- ◆ Stratified Random Sampling
 - Theory and data categories guide choices
- ◆ Cluster Sampling
 - Convenience and real life guide the selection

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Systematic Random Sample

- ◆ Every kth member is chosen
 - k is the "skip number" ($\text{Popsize/sample size} = N/n$)
- 1. Get list of cases (randomized to elim. bias)
- 2. Calculate K
 - round down or will run out of cases (e.g. for $N=100$ and $n=8$, use 12 *not* 13)
- 3. Randomly select 1 of 1st K cases in frame
 - use the table, but only for the start number
- ◆ Adv: faster, easier, cheaper, fewer mistakes
- ◆ Disadv: need complete sampling frame with no pattern (army, city blocks); beware periodicity; may be expensive

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Simple Random Sample

- ◆ Assumed in stat. theory (w/o replacement)
 - Each case has same chance of being selected, so
 - Each possible sample has the same chance of being selected (!!)

NOTE:

- Random is not the same as arbitrary
 - random means ignorance by choice and design
 - We don't know who we're picking

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Stratified Random Sampling

- ◆ Population divided into "strata" by variable(s)
 - groups that are similar on some characteristic
 - must be mutually exclusive (i.e. at least nominal)
- ◆ SiRS or SyRS from each stratum
- ◆ Could be proportional
 - Sample proportions chosen to reflect population
 - Black respondents, southern states,
 - Like quotas, but during sampling not observation
- ◆ Could also do disproportionate
 - Esp. when want to oversample a small group
- ◆ Adv: Sampling error is reduced by two factors
 - A large sample, and a homogeneous population
- ◆ Disadv: need adequate sampling frame
 - Typically, the entire population (*every* case)

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Simple Random Sample, cont'd

- ◆ Assign each case in sampling frame an ID #
- ◆ Use random number table to get sampled #s
 - First decide how you'll read it – which digits (first 2?) and which direction (down each column?)
 - Doesn't matter how you do it, just be consistent
 - Pick a random starting point (*just* the starting pt.)
 - Some numbers won't be useable; just skip those
- ◆ Adv: eliminates bias, permits inference
- ◆ Disadv: time consuming, need entire sampling frame, expensive (tho less w/ PCs)

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Disproportionate Strata

An over-sample of non-white....

10 Non-Whites (20%)

10 Whites (80%)

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Weighting & Sample Weights

- ◆ Used if sample is disproportionate
 - E.g. population is 12.8% African-American, but sample is 6.4% African-American
 - Use a data "trick": pretend there were twice as many African-Americans
 - "weight by race" in SPSS
- ◆ Something to remember & consider
- ◆ But you needn't weight data this term

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Multistage Design & Errors

- ◆ A two stage cluster sample is subject to two sampling errors, one at each stage.
 - Selecting a disproportionately wealthy city block plus disproportionately wealthy residents within those blocks.
 - ◆ Elements within a natural cluster are more homogeneous than are elements composing the total population
 - Birds of a feather flock together!

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Q3. For a highly disproportionate sample, it may be appropriate to...

- A. Cry 3%
- B. Dance 0%
- C. Give up 0%
- D. Start over 10%
- ➔ E. Weight the data

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87%



Clusters vs. Strata

- ◆ Clusters:
 - chosen for convenience and should be internally *heterogeneous*
- ◆ Strata:
 - dictated by theory and should be internally *homogenous*
- ◆ Example: MTV's *Real World*:
 - The cast of each season is a cluster
 - ◆ It's a practical group of possible cases
 - ◆ Any one cast is pretty much like the rest
 - ◆ Heterogenous within each
 - Strata would be groups across casts
 - ◆ E.g. age, race, gender, sexual orientation
 - ◆ Would need list of all the members of all the casts, or at least a list of percentages of each subgroup (male and female; black white and other, etc.)

Not in lecture
— notes only

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Cluster Sampling

- ◆ Steps:
 - Sampling frame is divided into clusters
 - ◆ Practical, natural groupings of cases rather than theoretical or demographic ways they differ
 - Randomly select clusters
 - Randomly select cases within those clusters
 - ◆ Sometimes use all cases within selected clusters
- ◆ Advantages
 - save time & money
 - just need list of clusters, not all cases
- ◆ Disadvantages:
 - may not have enough variation if small clusters
 - risk getting unusual clusters
 - ◆ biased if they are theoretically important (wealthy suburbs)
 - Statistical inference procedures very complicated
 - ◆ must people don't even use them

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4. Sampling theory assumes...

- ✓ A. Simple Random Sampling 43%
- B. Systematic Random Sampling 43%
- C. Cluster Sampling 10%
- D. Quota Sampling 0%
- E. Multi-Stage Sampling 3%

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Homogeneity vs. Heterogeneity

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- ◆ Failure of early polls was not adequately representing the full range of voters
- ◆ Sometimes ok to assume all cases alike
 - Blood samples
 - Social psychology - Many studies based on college sophomores in a psych class
- ◆ But usually, we're concerned with heterogenous populations
 - Indeed, we *focus* on differences & variations

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Considerations in Choosing N

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- ◆ Precision: *the width of the confidence interval*
- ◆ Confidence: *Probability that interval will contain population mean*
- ◆ Degree of Variability: *Use larger samples for more heterogeneous populations*
- ◆ Amount and Type of Analysis: *More complicated studies require larger samples*
- ◆ Resources: *How much research money do you have ...*

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What makes a good sample?

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- ◆ Ideal is to be *representative*
 - Aggregate characteristics clearly resemble same aggregate characteristics of the pop
 - Variation not *lacking* or *unbalanced*
 - In practice, may not *know* parameters
- ◆ Needn't be representative on *all* respects
 - Only on characteristics relevant to the study
 - But might not know what those are in advance
 - & Don't want criticism about hidden "control" variable
- ◆ Randomized selection increases *likelihood*
 - Does NOT guarantee it
 - Always a *chance* of sampling errors
 - Almost always *some* amount of error
 - Each *method* allows for particular errors

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Sample Size Required for Estimating Population Mean

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If we want the "margin of error" to be smaller than some value B, with a probability of at least $1 - \alpha$, the appropriate sample size is obtained as follows

$$n = \sigma^2 \left(\frac{z_{\alpha/2}}{B} \right)^2$$

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How big should my sample be?

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- ◆ Size matters, because it decreases error
 - But larger isn't always "better"
 - Pt. of diminishing returns
 - More important to be **unbiased**
 - Theoretical goal is representativeness
 - Best we can do is reduce barriers to that
- ◆ Best practice is to *reduce bias*
 - risks making sampling unrepresentative
 - However, any method introduces some bias
 - assumptions about availability of subjects
 - Assumes results approximately representative
- ◆ Probability sampling improves chances
 - by avoiding bias and allowing statistical inferences based on assumptions of random samples

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Admin Overview Non-Prob Probability Misc **Choosing** L33

Sample Size Required for Estimating Population Proportion

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If we want the "margin of error" to be smaller than some value B, with a probability of at least $1 - \alpha$, the appropriate sample size is obtained as follows

$$n = .25 \left(\frac{z_{\alpha/2}}{B} \right)^2$$

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What's a "good" sample size?

The formula for n is the *variance*, multiplied by the *square* of the ratio of the *z score* associated with the desired confidence level and the *allowed margin of error*:

$$n = \hat{\sigma}^2 * \left(\frac{z_{\alpha/2}}{B} \right)^2 = p * (1 - p) * \left(\frac{z_{\alpha/2}}{B} \right)^2$$

When the population proportions are unknown and we must use the conservative estimate of 50% (0.5), then to have a 95% level of confidence in a margin of error of not more than +/- 4% requires a sample size of 601 cases*:

$$0.5 * (0.5) * \left(\frac{1.96}{.04} \right)^2 = .25(49)^2 \approx 601$$

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* Per subgroup or disaggregated unit

5. What is my first name?

A. Babbie 0%

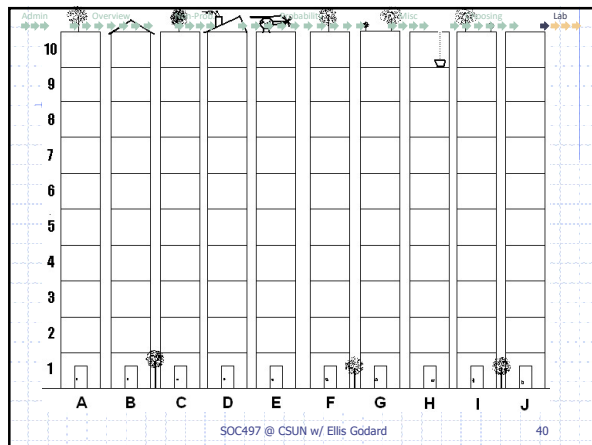
B. Donald 0%

✓ C. Ellis 97%

D. Godard 3%

E. Zebra 0%

00



Team Scores

Points	Team
4	Watching The Princes...
3.8	Weekend Zooms w/ Ellis
3.78	HW4
3.51	Studying for the midterm
3.5	"Catching up" on the...

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Today's Lab: Sampling Effects

- ◆ See 2-page handout of apartments
 - 10 buildings w/ 10 apts each, 1st floor to penthouse
- ◆ You'll do each (every!) form probability sampling
 - Drawing same sample size from same sampling frame
 - Compare results of the four different methods
- ◆ Numbering hints (not needed in A)
 - B: Use 00-99 (not 1-100), or else need 3-digit #s
 - C: 0-9 (per row, i.e. "floor" of the buildings), 10 times
 - D: 0-9 bldgs for SIRS, then 0-9 in each of 2 for SyRS

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