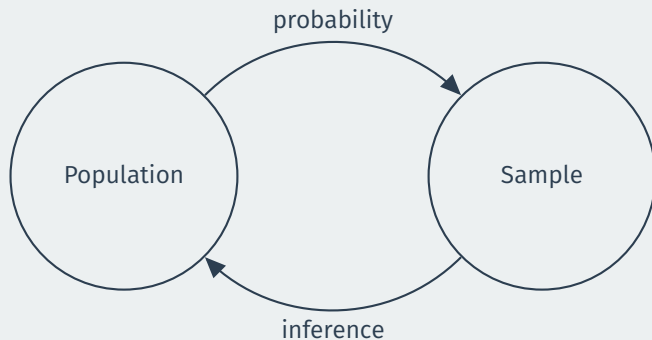


Gov 51: Estimators

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Remember our goal



- We want to learn about the chance process that generated our data.
- Now we switch to **inference**.
 - What can I learn about the population distribution from my sample?

How popular is Donald Trump?



- What proportion of the public approves of Trump's job as president?
- Latest Gallup poll:
 - Oct. 29th–Nov. 4th
 - 1500 adult Americans
 - Telephone interviews
 - Approve (40%), Disapprove (54%)
- What can we learn about Trump approval in the population from this one sample?

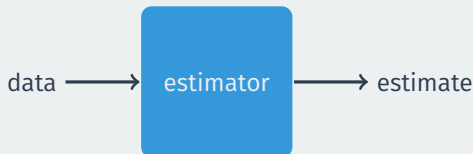
Samples from the population

- Simple random sample of size n from some population Y_1, \dots, Y_n
 - \rightsquigarrow i.i.d. random variables
 - e.g.: $Y_i = 1$ if i approves of Trump, $Y_i = 0$ otherwise.
- **Statistical inference:** using data to guess something about the population distribution of Y_i .

Point estimation

- **Quantity of interest:** some feature of the population distribution.
 - Also called: parameters.
 - These are the things we want to learn about.
- **Point estimation:** providing a single “best guess” about this q.o.i.
- Examples of quantities of interest:
 - $\mu = \mathbb{E}[Y_i]$: the population mean (turnout rate in the population).
 - $\sigma^2 = \mathbb{V}[Y_i]$: the population variance.
 - $\mu_1 - \mu_0 = \mathbb{E}[Y(1)] - \mathbb{E}[Y(0)]$: the population ATE.

Estimators



- **Estimator**: function of the data that produces estimates of the q.o.i.
 - An **estimate** is one particular realization of the estimator
- Ideally we'd like to know the **estimation error**, estimator — truth
 - Problem: θ is unknown.
- Solution: figure out the properties of estimator using probability.
 - Estimator is a r.v. because it is a function of r.v.s. (the data)
 - \rightsquigarrow estimator has a distribution has a distribution.

Estimating Trump's support

- Parameter p : **population proportion** of adults who support Trump
- There are many different possible estimators:
 - $\hat{p} = \bar{Y}_n$ the sample proportion of respondents who support Trump.
 - $\hat{p} = Y_1$ just use the first observation
 - $\hat{p} = \max(Y_1, \dots, Y_n)$
 - $\hat{p} = 0.5$ always guess 50% support
- How good are these different estimators?

- Assume a simple random sample of n voters: $n = 1500$
- Define r.v. Y_i for Trump approval:
 - $Y_i = 1 \rightsquigarrow$ respondent i approves of Trump
 - $Y_i = 0 \rightsquigarrow$ respondent i disapproves of Trump
- Y_i is **Bernoulli** with probability of success p
 - “success” = “selecting a Trump approver”
 - $p = \mathbb{P}(Y_i = 1)$ the population proportion of Trump approvers.
- Sample proportion is the same as the sample mean:

$$\bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i = \frac{\text{number who support Trump}}{n}$$

Sample mean properties

sample proportion = population proportion + chance error

$$\bar{Y} = p + \text{chance error}$$

- Remember: the sample mean/proportion is a random variable.
 - Different samples give different sample means.
 - Chance error “bumps” sample mean away from population mean
- $\rightsquigarrow \bar{Y}$ has a distribution across repeated samples.

Central tendency of the sample mean

- Expectation: average of the estimates across repeated samples.
 - From last week, $\mathbb{E}[\bar{Y}] = \mathbb{E}[Y_i] = p$
 - \rightsquigarrow chance error is 0 on average:

$$\mathbb{E}[\bar{Y} - p] = \mathbb{E}[\bar{Y}] - p = 0$$

- **Unbiasedness:** Sample proportion is on average equal to the population proportion.

Spread of the sample mean

- **Standard error:** how big is the chance error on average?
 - This is the standard deviation of the estimator.
- Special rule for sample proportions:

$$\sqrt{\mathbb{V}(\bar{Y})} = \sqrt{\frac{p(1-p)}{n}}$$

- Problem: we don't know p !
- Solution: **estimate** the SE:

$$\sqrt{\hat{\mathbb{V}}(\bar{Y})} = \sqrt{\frac{\bar{Y}(1-\bar{Y})}{n}} = \sqrt{\frac{0.37 \times (1-0.37)}{1500}} \approx 0.012$$