INTRODUCTION TO DATA ANALYSIS

Data Analysis for Journalism and Political Communication (Spring 2025)

Prof. Bell

SMPA 2152 Introduction Prof. Bell 1/9

EXPECTATIONS

- An A effort: Starts homework early and asks questions/attends office hours; reviews the professor's feedback; reads assignment instructions and grading rubrics carefully; attends class; being perfect at coding is not an expectation of an A effort.
- A B effort: Completes all homework assignments; does not carefully read assignment instructions; uses outside resources to complete the homework; attends class; a B effort meets minimum expectations.
- A C effort or lower: Does not complete all homework assignments; does not carefully read assignment instructions; does not communicate with the professor; does not consistently attend class; I will reach out if you meet these criteria.

2/9

WHEN YOU GET STUCK

- Look back at the lecture materials. Everything required to complete the homework is in the lecture materials.
- ② Google the error (this is different than Googling "how do I do X?").
- Semail the professor. Be sure to include enough code in your email that I can recreate the problem.

CHALABI: 3 WAYS TO SPOT A BAD STATISTIC



4/9

SMPA 2152 Introduction Prof. Bell

CHALABI: 3 WAYS TO SPOT A BAD STATISTIC



- Can you see uncertainty?
- 2 Can we look beyond the averages?
- Mow was the data collected?



SMPA 2152

We rarely get a complete count of everything



5/9

SMPA 2152 Introduction Prof. Bell

- We rarely get a complete count of everything
- How do we know that the **sample** we choose is a good representative of the whole **population**?

- We rarely get a complete count of everything
- How do we know that the **sample** we choose is a good representative of the whole **population**?
- We will learn how to measure and communicate about uncertainty, which is both art and science

5/9

- We rarely get a complete count of everything
- How do we know that the sample we choose is a good representative of the whole **population**?
- We will learn how to measure and communicate about uncertainty, which is both art and science
- Humans do not do well with probability

- We rarely get a complete count of everything
- How do we know that the **sample** we choose is a good representative of the whole **population**?
- We will learn how to measure and communicate about uncertainty, which is both art and science
- Humans do not do well with probability

"There are only five probabilities the average human can handle: 99 percent, one percent, 100 percent, zero, and 50-50. That's it."

- Richard Thaler, Nobel Laureate in Economics

- We rarely get a complete count of everything
- How do we know that the **sample** we choose is a good representative of the whole **population**?
- We will learn how to measure and communicate about uncertainty, which is both art and science
- Humans do not do well with probability

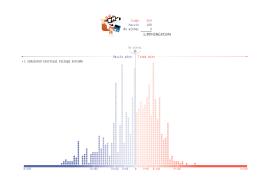


5/9

- We rarely get a complete count of everything
- How do we know that the sample we choose is a good representative of the whole population?
- We will learn how to measure and communicate about uncertainty, which is both art and science
- Humans do not do well with probability

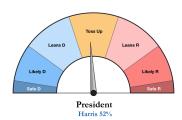


- We rarely get a complete count of everything
- How do we know that the **sample** we choose is a good representative of the whole **population**?
- We will learn how to measure and communicate about uncertainty, which is both art and science
- Humans do not do well with probability



- We rarely get a complete count of everything
- How do we know that the **sample** we choose is a good representative of the whole **population**?
- We will learn how to measure and communicate about uncertainty, which is both art and science
- Humans do not do well with probability

2024 Presidential Forecast



241 Harris, 246 Trump, 51 Toss Up

 There is always a trade-off between accuracy and simplicity when working with data

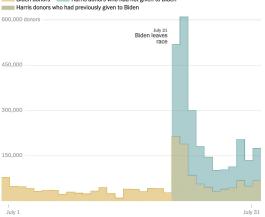


- There is always a trade-off between accuracy and simplicity when working with data
- We aggregate data to make it easier to comprehend, but we may also lose important context

Donors both old and new gave to the newly renamed Harris campaign

Both donors who had given to the Biden re-election campaign and new people who had not previously contributed rushed to donate to the Harris campaign.

Biden donors Harris donors who had not given to Biden



Source: Federal Election Commission - The New York Times

- There is always a trade-off between accuracy and simplicity when working with data
- We aggregate data to make it easier to comprehend, but we may also lose important context
- We will talk about using data visualization to communicate about data, as well as researcher choices and biases

- There is always a trade-off between accuracy and simplicity when working with data
- We aggregate data to make it easier to comprehend, but we may also lose important context
- We will talk about using data visualization to communicate about data, as well as researcher choices and biases
- We will also talk about the importance of theory in understanding data, especially correlation vs. causation

Storks Deliver Babies (p = 0.0	008)	
EXYMORUS. Teaching: Symplectic control of the contr	Robert Matthews Aston University, Birmingham, England. e-mail: rajm@compuserve.com	
	Summary This article shows that a highly statistically significant correlation exists between stork populations and human birth rates across Europe. While stocks may not deliver babies, unthinking interpretation of correlation and p-values can certainly deliver unreliable conclusions.	
◆ INTRODUCTION ◆	association between storks and the concept of women as bringers of life, and also in the bird feeding labbits, which were conce regarded as search for embryonic life in water (Cooper 1922. The logend lives on to this day, with necessi- basting storks being a regular feature of greeting cards celebrating births.	
Introductory statistics textbooks routinely warn of the dangers of confusing correlation with causation, pointing out that while a high correlation coefficient is indicative of (linear) association, it cannot be taken as a measure of causation. Such		

• Data is not objective – it is generated by humans



7/9

SMPA 2152 Introduction Prof. Bell

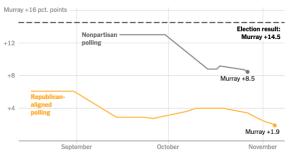
HOW WAS THE DATA COLLECTED?

- Data is not objective it is generated by humans
- Some data is produced by unscrupulous actors



SMPA 2152 Introduction

- Data is not objective it is generated by humans
- Some data is produced by unscrupulous actors

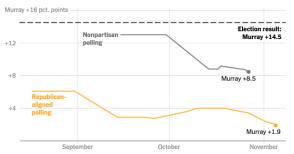


Source: New York Times analysis of Washington Senate race polls aggregated by FiveThirtyEight • Notes: Trends are calculated with a 14-day average. Polling groups considered Republican-aligned include those identified by The New York Times and FiveThirtyEight. Polling groups considered nonpartisan are those not known to be aligned with or funded by a political party. • By Jason Kao

SMPA 2152 Introduction Prof. Bell 7/9

HOW WAS THE DATA COLLECTED?

- Data is not objective it is generated by humans
- Some data is produced by unscrupulous actors



Source: New York Times analysis of Washington Senate race polls aggregated by FiveThirtyEight • Notes: Trends are calculated with a 14-day average. Polling groups considered Republican-aligned include those identified by The New York Times and FiveThirtyEight. Polling groups considered nonpartisan are those not known to be aligned with or funded by a political party. • By Jason Kao

 But most of the time, poor analysis is not nefarious – humans are imperfect

 Garbage in = garbage out: no amount of statistical wizardry can compensate for bad data



7/9

SMPA 2152 Introduction Prof. Bell

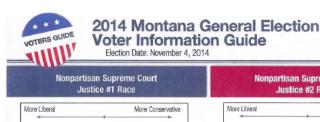
- Garbage in = garbage out: no amount of statistical wizardry can compensate for bad data
- We will spend a lot of time thinking about the data generating process and how it can bias our results

Support for mass deportation varies depending on how the question is asked

A sample of different questions asked about deportation this year show significant variation in levels of support — sometimes, even within the same survey

Pollster	Wording	Support deportations v	Do not support deportations
CBS News/YouGov Registered voters, June 5-7	Would you favor or oppose the U.S. government starting a new national program to deport all undocumented immigrants currently living in the U.S. lilegally?	62%	38%
Marquette Law School Registered voters, October 1-10	Do you favor or oppose deporting immigrants who are living in the United States illegally back to their home countries?	58%	42%
ABC/Ipsos US adults, October 4-8	There are at least 11 million undocumented immigrants living in the United States. Would you support or oppose an effort by the federal government to deport all these undocumented immigrants and send them back to their home countries?	56%	43%
CNN US adults, January 25-30	If Donald Trump becomes president again, would you favor or oppose him trying todetain and deport millions of undocumented immigrants?	48%	52%
Gallup US adults, June 3-23	Please tell me whether you strongly favor, favor, oppose, or strongly oppose each of the following proposalsDeporting all immigrants who are living in the United States illegally back to their home country	47%	51%
Marquette Law School Registered voters, October 1-10	Do you favor or oppose deporting immigrants who are living in the United States illegally back to their home countries even if they have lived here for a number of years, have jobs and no criminal record?	40%	60%
Pew Research US adults, April 8-14	Which comes closer to your view about how to handle undocumented immigrants who are now living in the U.S.? (They should not be allowed to stay in the country legally. There should be a way for them to stay in the country legally, if certain requirements are met) If "not be allowed: "Do you think there should be a national law	33%	67%
мрті в-24	enforcement effort to deport all immigrants who are now living in the U.S. lilegally?" ("Support" percentage includes those who say there should be a national deportation effort, "do not support" includes all others)		

- Garbage in = garbage out: no amount of statistical wizardry can compensate for bad data
- We will spend a lot of time thinking about the data generating process and how it can bias our results
- We will also discuss our ethical responsibilities around data



Mitt Romney

W. David Herbert

Nonpartisan Supreme Court Justice #2 Race



For more information on how these figures were created, please see http://data.stanford.edu/dime. Please note that this guide is non-partisan and does not endorse any candidate or party. This guide was created as part of a joint research project at Stanford and Dartmouth.

Paid for by researchers at Stanford University and Dartmouth College, 616 Serra Street, Stanford, CA 94305

Jim Rice

Barack Ohama

Take this to the polls!

7/9

SMPA 2152 Prof Bell

GROUP DISCUSSION

Introduce yourself to your neighbor(s) and take a few minutes to review these additional graphs from Mona Chalabi. Do any of these stand out to you as being good (or bad) examples of our three questions for spotting a bad statistic?

- Can you see uncertainty?
- ② Can we look beyond the averages?
- Mow was the data collected?

On your notecard, please write:

- Preferred name
- Preferred pronouns
- Year in school and major
- Your background in coding and/or statistics
- One thing you hope to get out of this class

9/9