

# gesis

Leibniz Institute  
for the Social Sciences



## Potentials and Pitfalls of Social Media Data

*Indira Sen & Katrin Weller*  
GESIS workshop - December 2022

# Hands-on Error Analysis

- Python notebook
- Stance vs sentiment
- How accurate is VADER?
- Data collection issues based on different subreddits

# How do we measure approval?

O'Connor, B., Balasubramanyan, R., Routledge, B.R. and Smith, N.A., 2010, May. **From tweets to polls: Linking text sentiment to public opinion time series.** In *Fourth international AAAI conference on weblogs and social media*.

Conrad, F.G., Gagnon-Bartsch, J.A., Ferg, R.A., Schober, M.F., Pasek, J. and Hou, E., 2019. **Social media as an alternative to surveys of opinions about the economy.** *Social Science Computer Review*, p.0894439319875692.

hate	negative
honest	positive
inefficient	negative
love	positive
destroy	negative
encourage	positive
...	...

# Are we Measuring approval towards Trump?

- Usually approval is defined as ‘sentiment’

<b>Tweet</b>	<b>Untargeted Sentiment</b>	<b>Approval</b>
Trump is the only candidate I fully support	positive	approval

# Are we Measuring approval towards Trump?

- This sentiment is untargeted, therefore, if multiple users are mentioned, untargeted sentiment  $\neq$  approval

<b>Tweet</b>	<b>Untargeted Sentiment</b>	<b>Targeted Sentiment</b>	<b>Approval</b>
Trump is the only candidate I fully support	positive	positive	approval
What makes me angry is the media unnecessarily attacking Trump	negative	positive	approval

# Are we Measuring approval towards Trump?

- Finally, targeted sentiment doesn't work when users are ***not directly*** mentioned

Tweet	Untargeted Sentiment	Targeted Sentiment	Stance	Approval
Trump is the only candidate I fully support	positive	positive	favor	approval
What makes me angry is the media unnecessarily attacking Trump	negative	positive	favor	approval
Jeb Bush is the best choice in the republican lineup	positive	none	against	disapproval

# What is ‘Stance’ Detection?

- “the task of automatically determining from text whether the author of the text is in favor of, against, or neutral towards a proposition or target.” [Mohammad et al. 2017]
- **Indirect stance:** “the target is referred to in indirect ways such as through pronouns, epithets, honorifics, and relationships.”

# Stance as a Theoretical Proxy for Approval

Tweet	Untargeted Sentiment	Targeted Sentiment	Stance	Approval
Trump is the only candidate I fully support	positive	positive	favor	approval
What makes me angry is the media unnecessarily attacking Trump	negative	positive	favor	approval
Jeb Bush is the best choice in the republican lineup	positive	none	against	disapproval



# Agenda

- Session 1: Introduction to Research with Social Media Data (SMD)
- Session 2: SM Data Collection
- Session 3: SMD Preprocessing and Analysis
- Session 4: Potential Pitfalls of SMD
- Session 5: Identifying Pitfalls with help from surveys
- Session 6: Identifying Pitfalls in SMD
- **Session 7: Mitigating Pitfalls**
- Session 8: Documenting Pitfalls
- Session 9: Recap and Conclusions

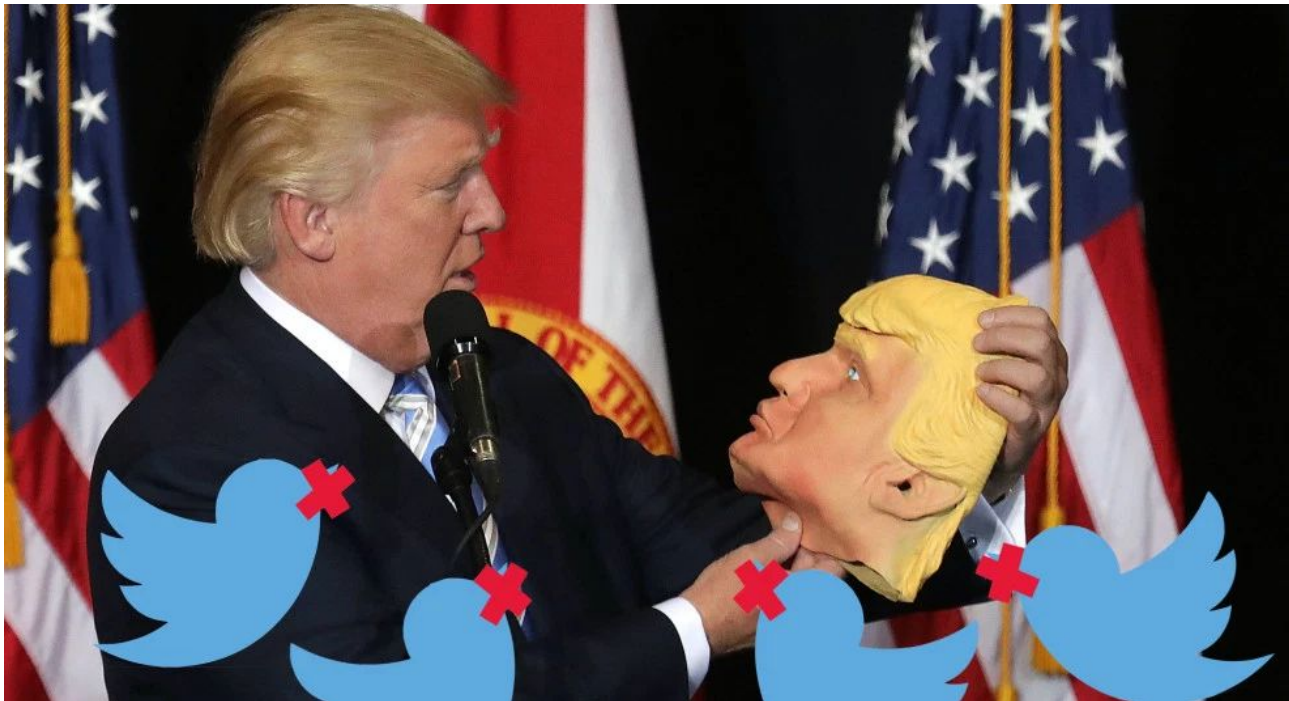
## 5: Mitigating Pitfalls





# Case Study

Understanding **Presidential Approval** from tweets



- Understand



## Who's Tweeting About the President? What Big Survey Data Can Tell Us About Digital Traces?

Josh Pasek<sup>1</sup>, Colleen A. McClain<sup>1</sup>, Frank Newport<sup>2</sup>,  
and Stephanie Marken<sup>2</sup>

Social Science Computer Review  
1-18  
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## Less is more? How demographic sample weights can improve public opinion estimates based on Twitter data.

**Pablo Barberá**  
Center for Data Science  
New York University  
[www.pablobarbera.com](http://www.pablobarbera.com)

## From Tweets to Polls: Linking Text Sentiment to Public Opinion Time Series

**Brendan O'Connor<sup>†</sup>** **Ramnath Balasubramanyan<sup>†</sup>** **Bryan R. Routledge<sup>§</sup>** **Noah A. Smith<sup>†</sup>**  
[brenocon@cs.cmu.edu](mailto:brenocon@cs.cmu.edu) [rbalasub@cs.cmu.edu](mailto:rbalasub@cs.cmu.edu) [routledge@cmu.edu](mailto:routledge@cmu.edu) [nasmith@cs.cmu.edu](mailto:nasmith@cs.cmu.edu)

<sup>†</sup>School of Computer Science  
Carnegie Mellon University

<sup>§</sup>Tepper School of Business  
Carnegie Mellon University

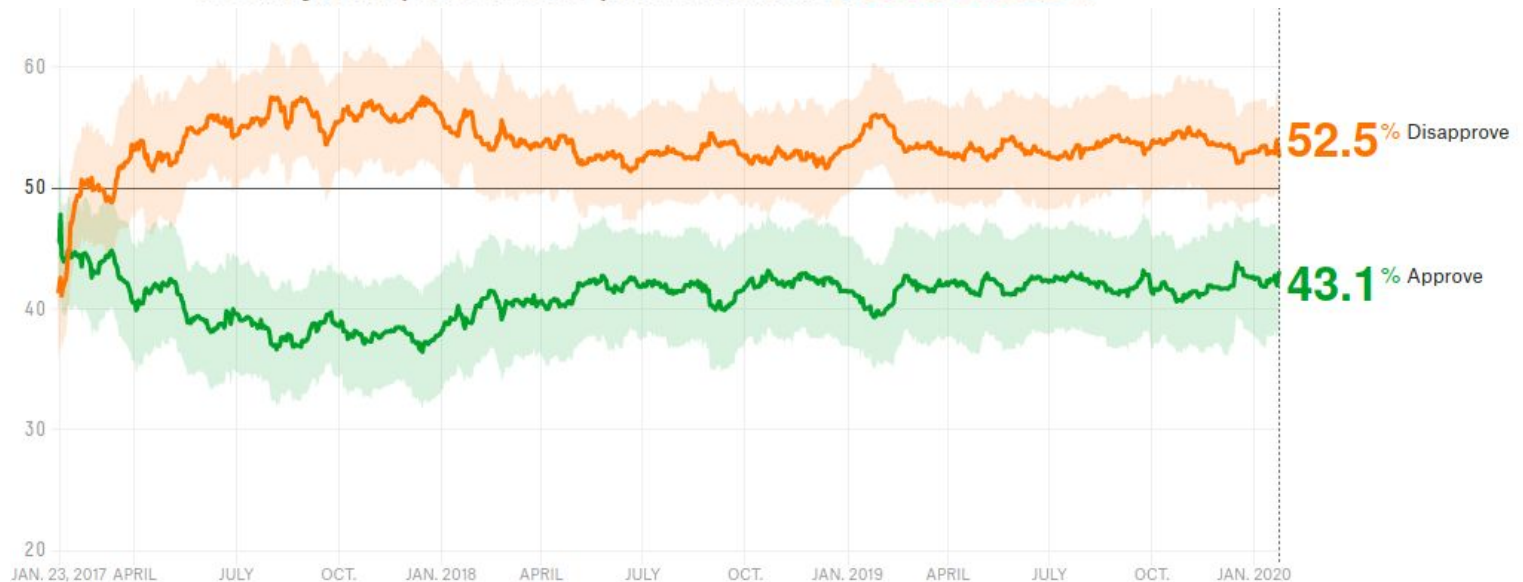
# Presidential Approval in Surveys

Do you approve or disapprove of the way Donald Trump is handling his job as president?

UPDATED JAN. 24, 2020 AT 2:15 PM

## How popular is Donald Trump?

An updating calculation of the president's approval rating, accounting for each poll's quality, recency, sample size and partisan lean. [How this works »](#)



**Data  
Analysis**

**Construct  
Definition**

Tweets about Trump with negative  
or positive sentiment

TRACES

Replying to

I was one that voted for Trump...BIGGEST mistake. I've  
always been blue but I thought I was voting for the  
Thank you 🇷🇺

Data  
Collection

Data  
Preprocessing

Data  
Analysis



Validity

Construct  
Definition

Tweets about Trump with negative  
or positive sentiment

TRACES

Replying to

I was one that voted for Trump...BIGGEST mistake. I've  
always been blue but I thought I was voting for the  
Thank you 🇷🇺

Data  
Collection

Tweets about Trump maybe related  
to his personal life

Data  
Preprocessing



Do you approve or disapprove of the way Donald Trump is handling his job as president?

Data  
Analysis



## MEASUREMENT

Construct  
Definition

Validity

Platform  
Selection

### Potential remedies for validity issues:

Build on a strong theoretical foundation for the expected traces

- Conduct focus groups with Twitter users and learn how they express approval
- Online panels: combine platform data with offline records (potentially very sensitive)

- Joseph, Kenneth, et al. "Polarized, Together: Comparing Partisan Support for Trump's Tweets Using Survey and Platform-Based Measures." Proceedings of the *International AAAI Conference on Web and Social Media*. Vol. 13. No. 01. (2019).
- Barberá, Pablo. "Less is more? How demographic sample weights can improve public opinion estimates based on Twitter data." *Work Paper NYU* (2016).
- Pew Research on how people tweet: <https://pewrsr.ch/2SgbujQ>





3 · Trending  
**#IEndorseBernie**  
3,700 Tweets



**Platform  
Affordances  
Error**



3 · Trending  
**#IEndorseBernie**  
3,700 Tweets



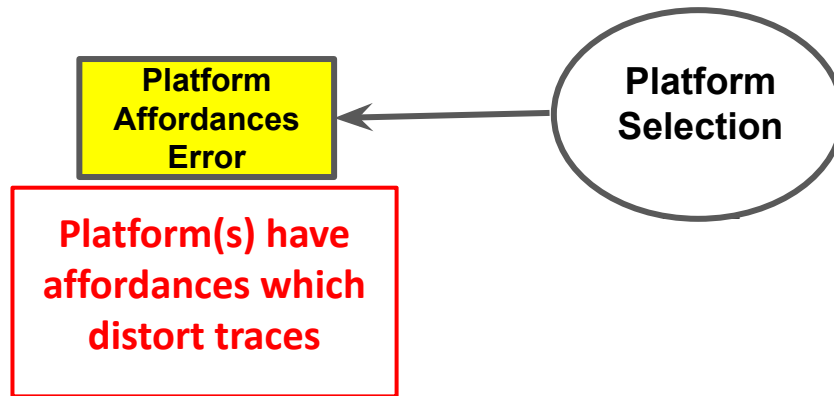
### Platform affordances:

- Terse tweets or a thread consisting of multiple tweets to express their opinion
- Less likely to expose an opinion that they assume to be unpopular on the platform
- Twitter’s “trends” feature, showing popular (including political) hashtags might nudge users

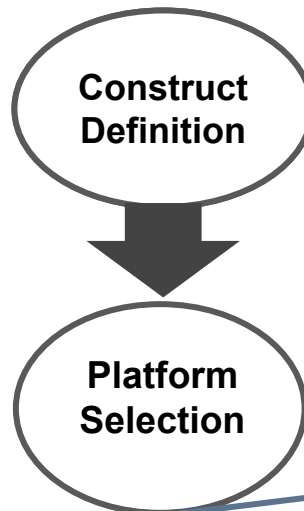
## MEASUREMENT

### Potential remedies for Platform Affordances Error:

- Quantify the effect of a feature vs. its non-existence
- Technical solutions around trace sharing restrictions



- Malik, Momin M., and Jürgen Pfeffer. "Identifying platform effects in social media data." *Tenth International AAAI Conference on Web and Social Media* (2016).
- Gligorić, Kristina, Ashton Anderson, and Robert West. "How constraints affect content: The case of Twitter's switch from 140 to 280 characters." *Twelfth International AAAI Conference on Web and Social Media* (2018).



Platform  
Coverage Error



Platform population = people with  
Twitter accounts != general American  
population

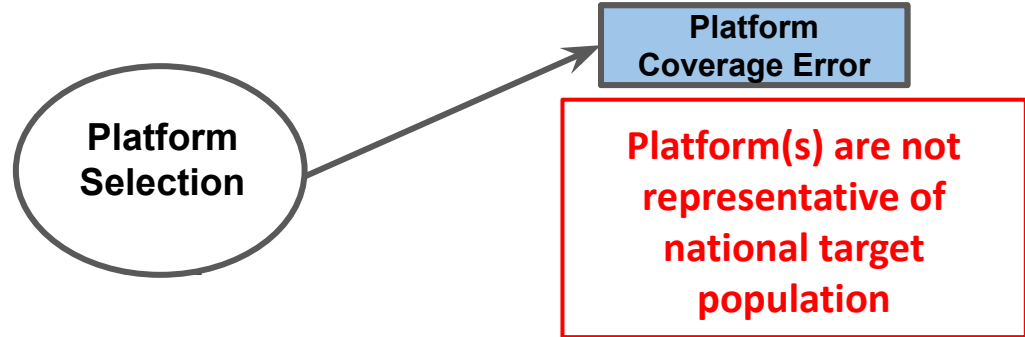
APRIL 24, 2019

## Sizing Up Twitter Users

*U.S. adult Twitter users are younger and more likely to be Democrats than the general public. Most users rarely tweet, but the most prolific 10% create 80% of tweets from adult U.S. users*

BY STEFAN WOJCIK AND ADAM HUGHES

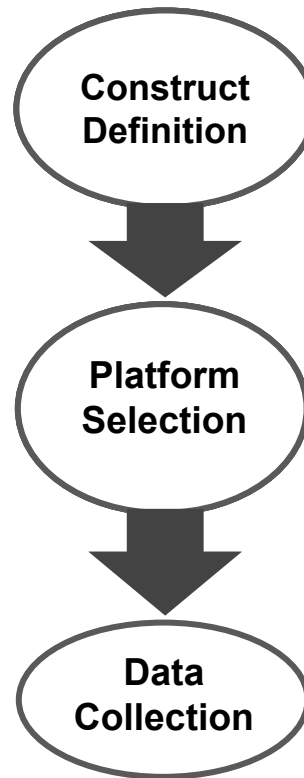
Data  
Analysis



**Potential remedies for Platform Coverage Error:**

- **Reweighting (adjustment) of platform population**
  - **Using multiple platforms**

→ Zagheni, Emilio, and Ingmar Weber. "Demographic Research with Non-Representative Internet Data." (2014).



**Trace  
Selection Error**

Tweets with keywords related to Trump:

- ‘Trump’ phrases unrelated to Donald Trump, which e.g. refer to Melania Trump or Donald Trump Jr.
- Likewise, relevant queries might be excluded that simply refer to “the president”, “Donald” or acronyms.



## MEASUREMENT

Construct  
Definition

### Potential remedies for Trace Selection Error:

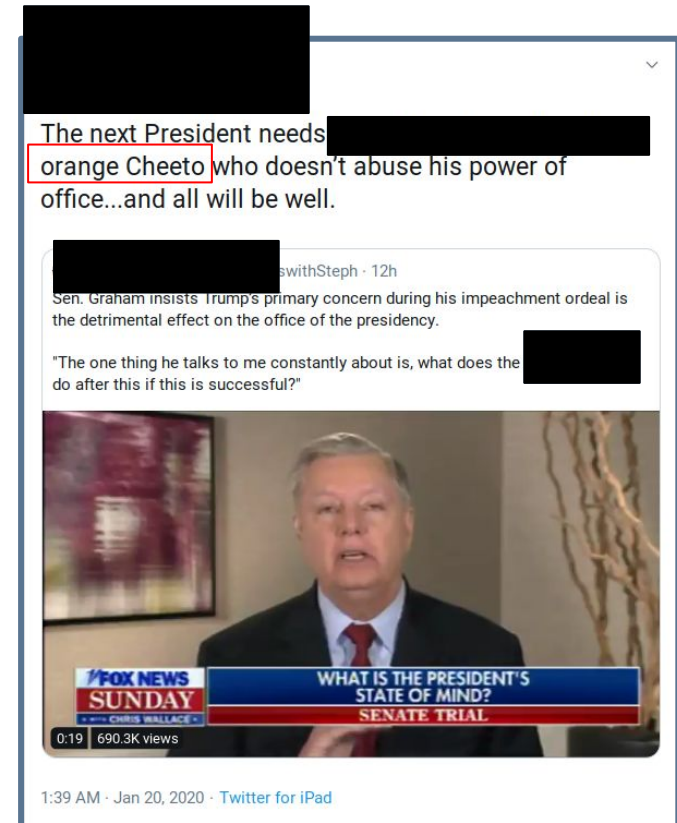
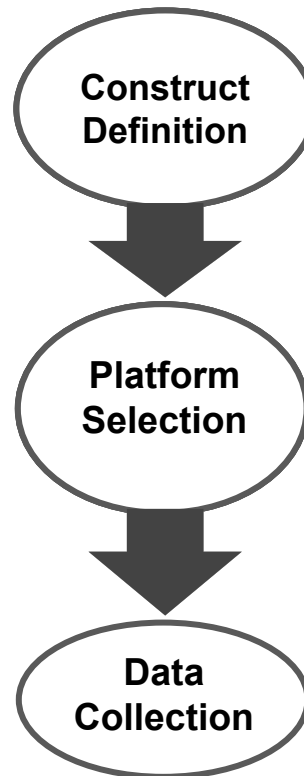
- Error analysis of queries, specifically understanding precision and recall
- Improving data collection strategies with query expansion

Trace  
Selection Error

Data  
Collection

Data  
Preprocessing

- Ruiz, Eduardo J., Vagelis Hristidis, and Panagiotis G. Ipeirotis. "Efficient filtering on hidden document streams." *Eighth International AAAI Conference on Weblogs and Social Media* (2014).
- Linder, Fridolin. "Improved data collection from online sources using query expansion and active learning." Available at SSRN 3026393 (2017).



TOP DEFINITION

# orange cheeto

A [millionaire](#) president with a [cheap ass tan](#)

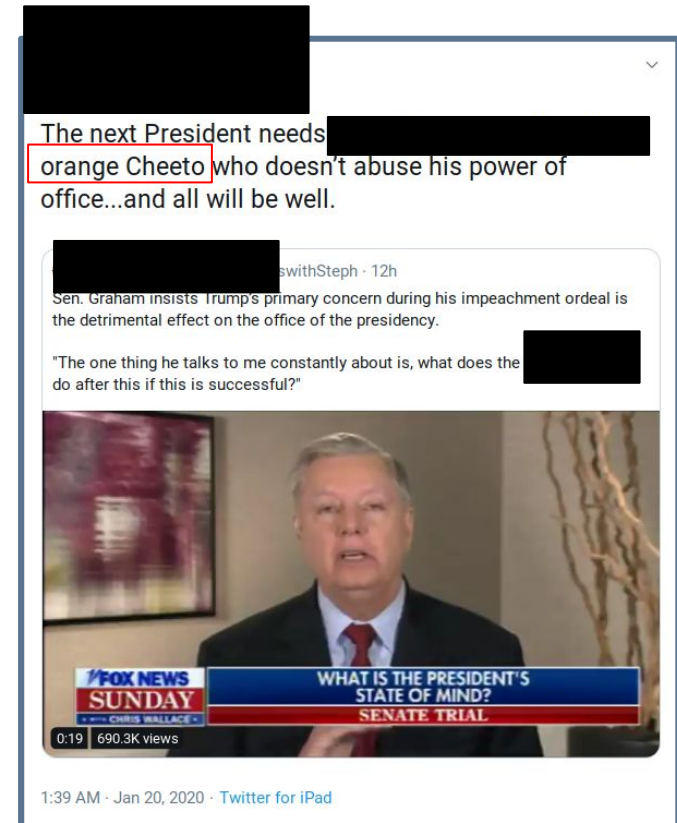
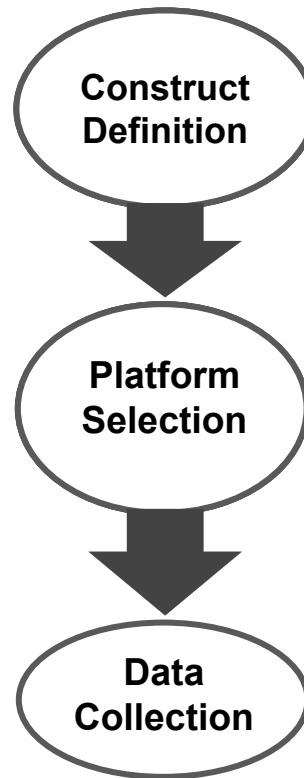
*\*hits blunt\** is that an orange [Cheeto](#)

[Nah dude](#) it's our 45th president

by [Brownperson1256](#) November 16, 2017







**User Selection  
Error**

Certain groups of users may be underrepresented if they refer to the President with a certain nickname or other name: 'orange cheeto' for detractors

Construct  
Definition

### Potential remedies for User Selection Error:

- Assess the characteristics of users included in the data with changing queries
- Reweighting (adjustment)

Data  
Collection

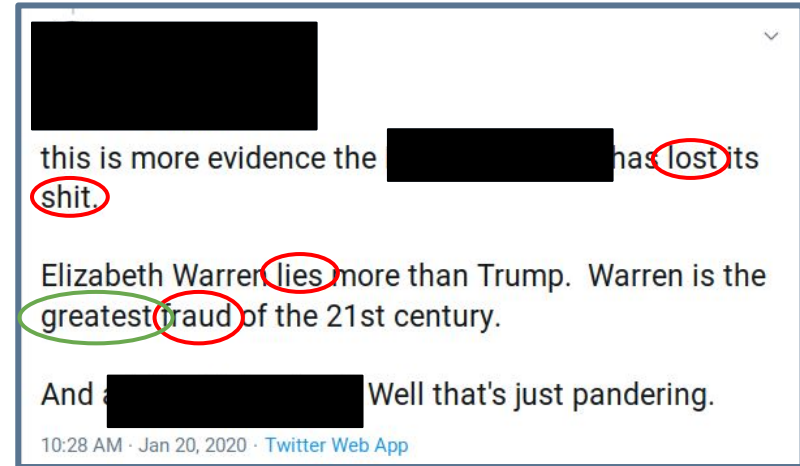
User Selection  
Error

Excluding Spanish  
tweets excludes  
Hispanic population of  
the US. Alternately,  
includes English  
speakers from other  
countries

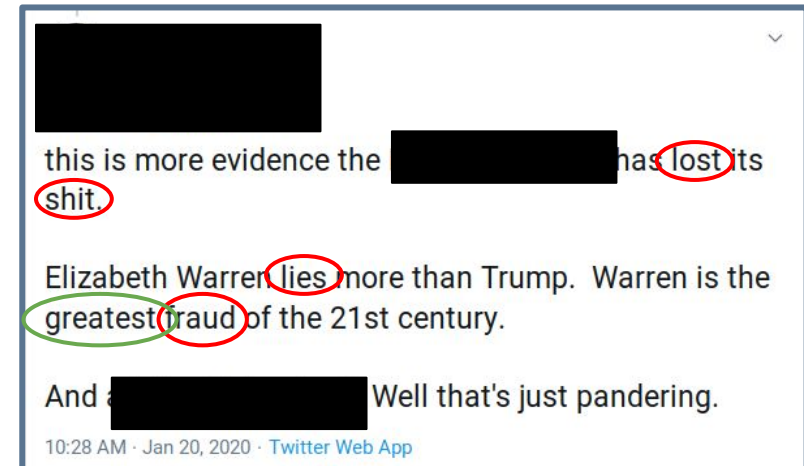
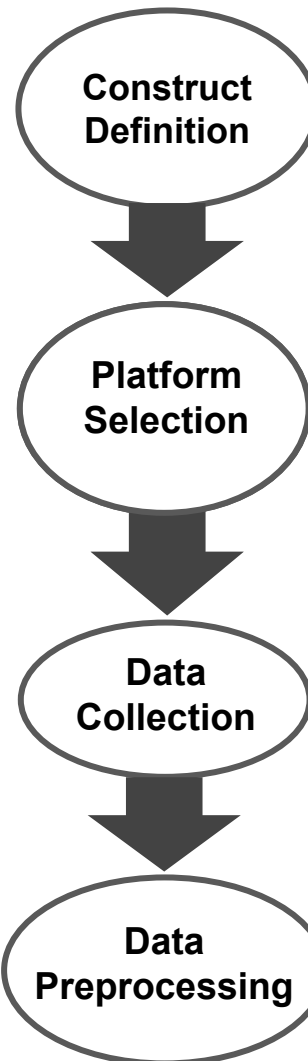
Data  
Preprocessing

Data  
Analysis

- Barberá, Pablo. "Less is more? How demographic sample weights can improve public opinion estimates based on Twitter data." *Work Paper NYU* (2016).



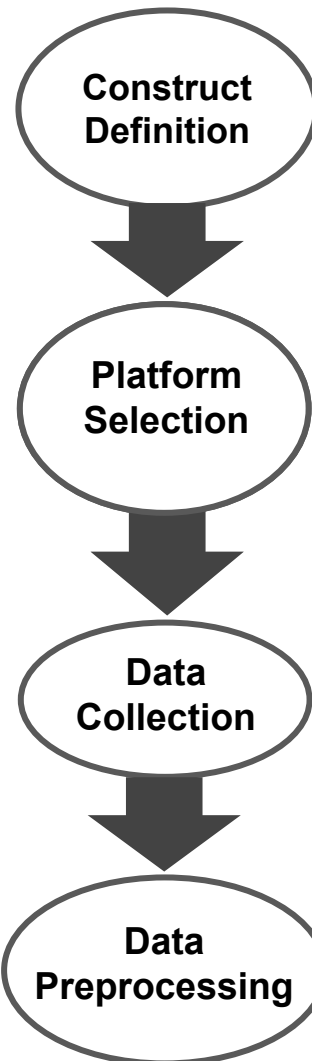
Trace  
Augmentation  
Error



**Augmentation:** sentiment analysis for annotating approval --- social media vocabulary mismatch, target-independent lexicon



Trace  
Reduction  
Error



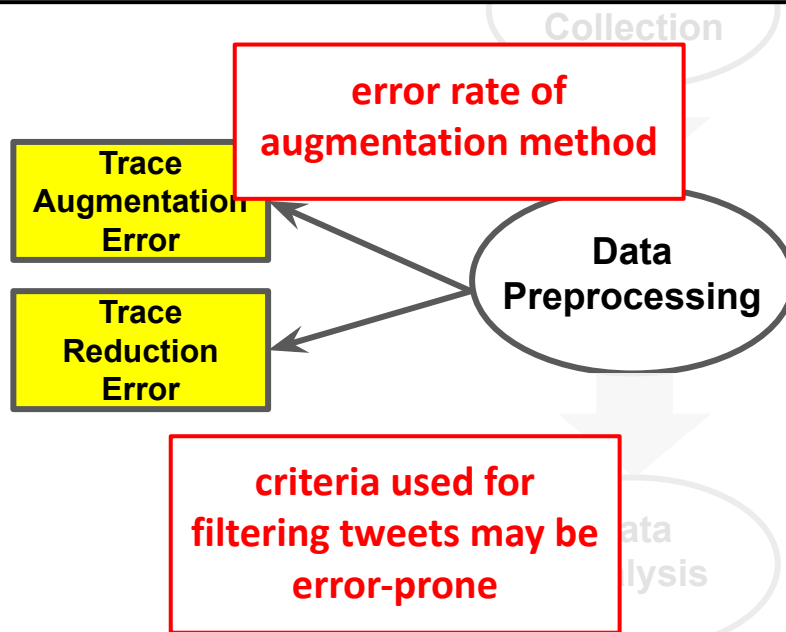
**Reduction:** remove spam tweets --- may inadvertently remove useful traces or discard tweets that do not contain any textual content, thereby ignoring relevant statements made through hyperlinks or embedded pictures/videos

## MEASUREMENT

- Denny, Matthew J., and Arthur Spirling. "Text preprocessing for unsupervised learning: why it matters, when it misleads, and what to do about it." *Political Analysis* 26.2 (2018): 168-189.
- González-Bailón, Sandra, and Georgios Paltoglou. "Traces of public opinion in online communication: A comparison of methods and data sources." *The ANNALS of the American Academy of Political and Social Science* 659.1 (2015): 95-107.

### Potential remedies for Trace Preprocessing Error:

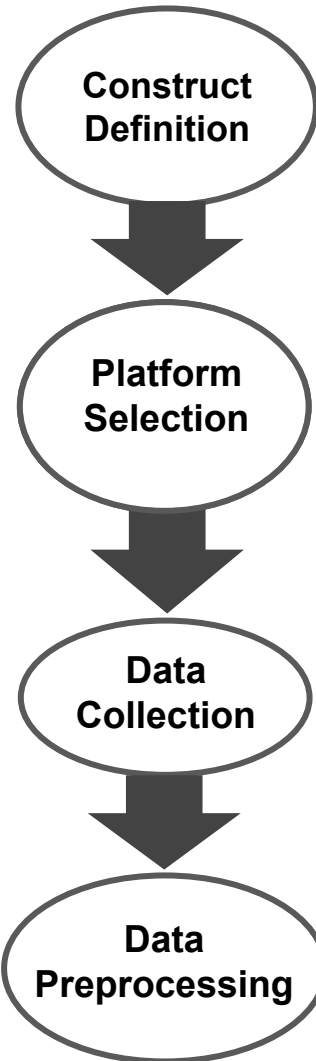
- Assess which augmentation and reduction methods are suitable





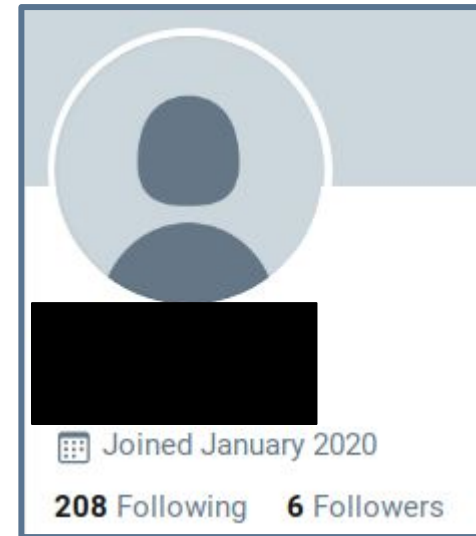


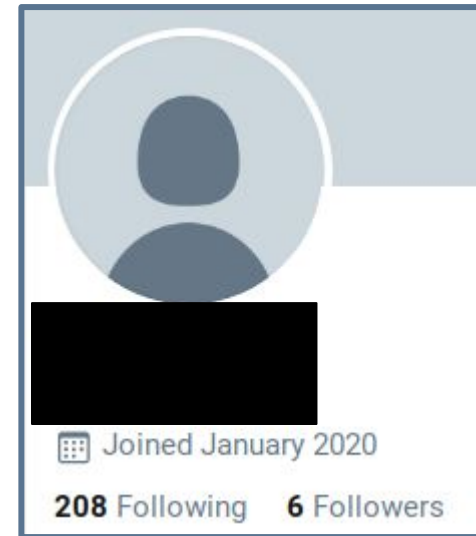
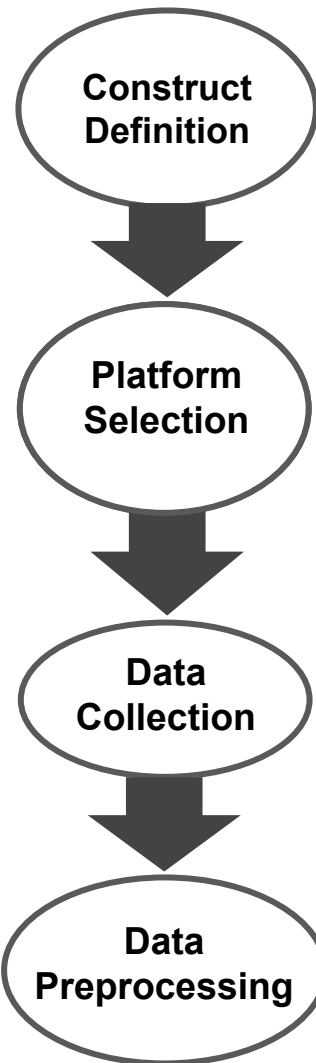




User  
Augmentation  
Error

**Augmentation:** Twitter profiles' age, gender, ethnicity, or location may be inferred to understand how presidential approval differs across demographics. But current methods have limitations

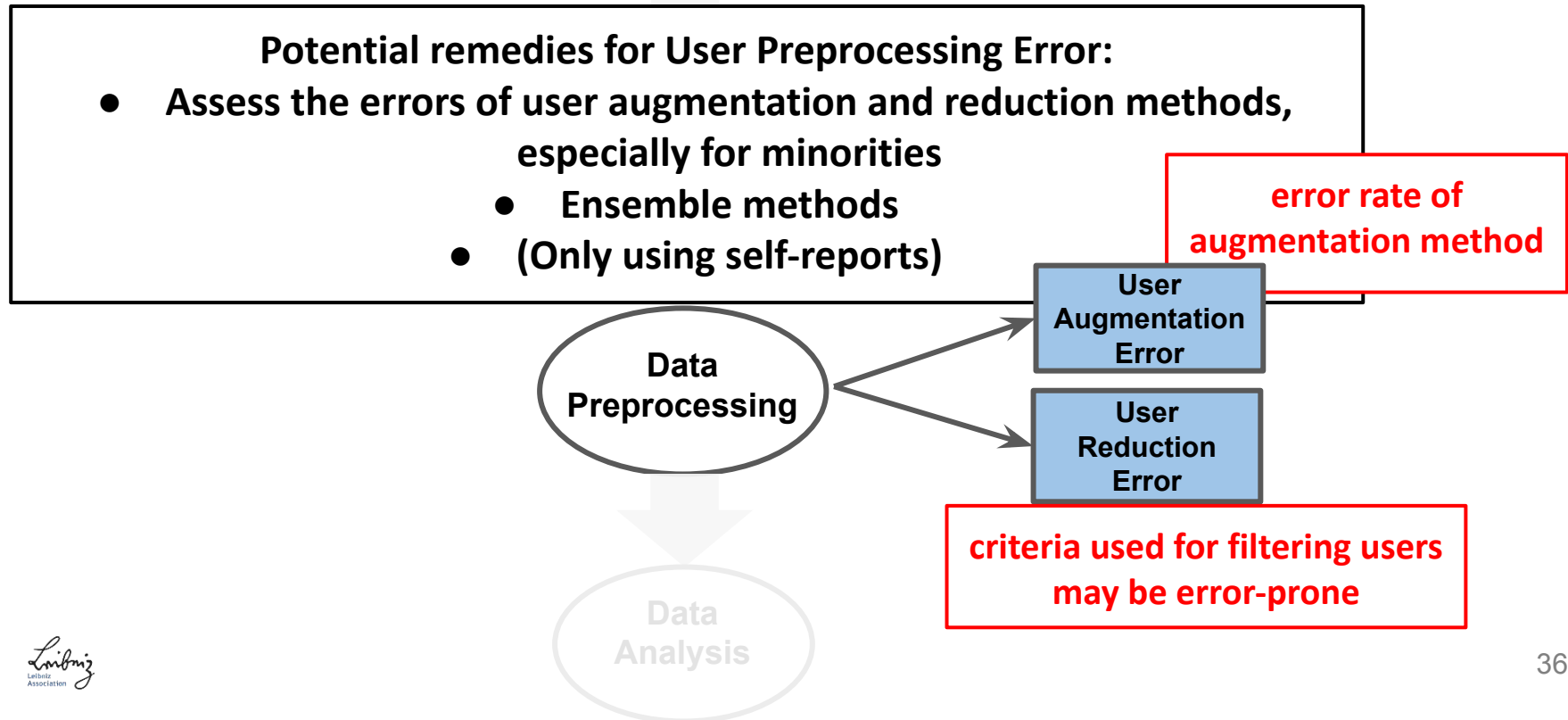


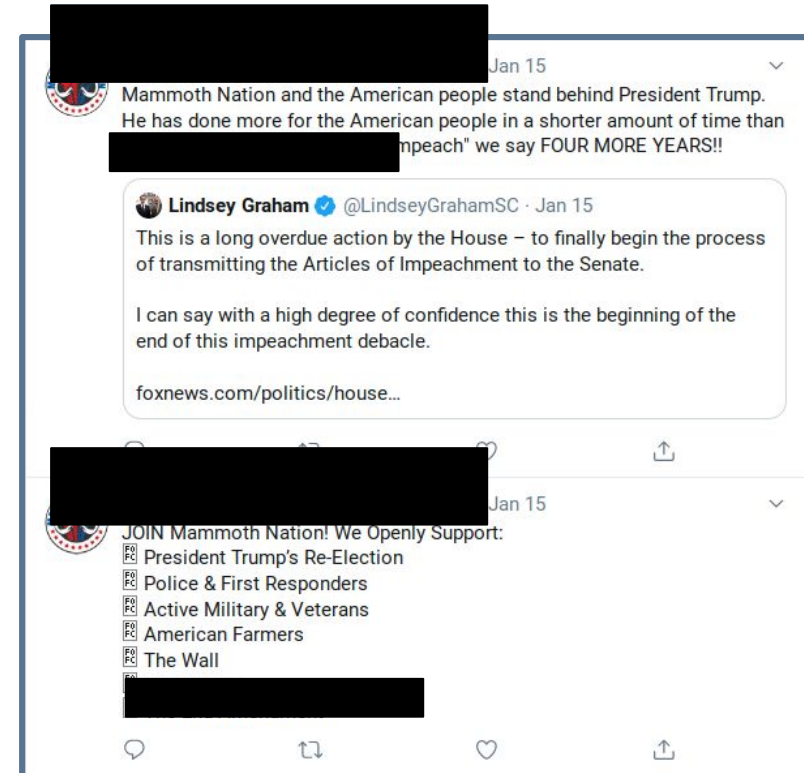


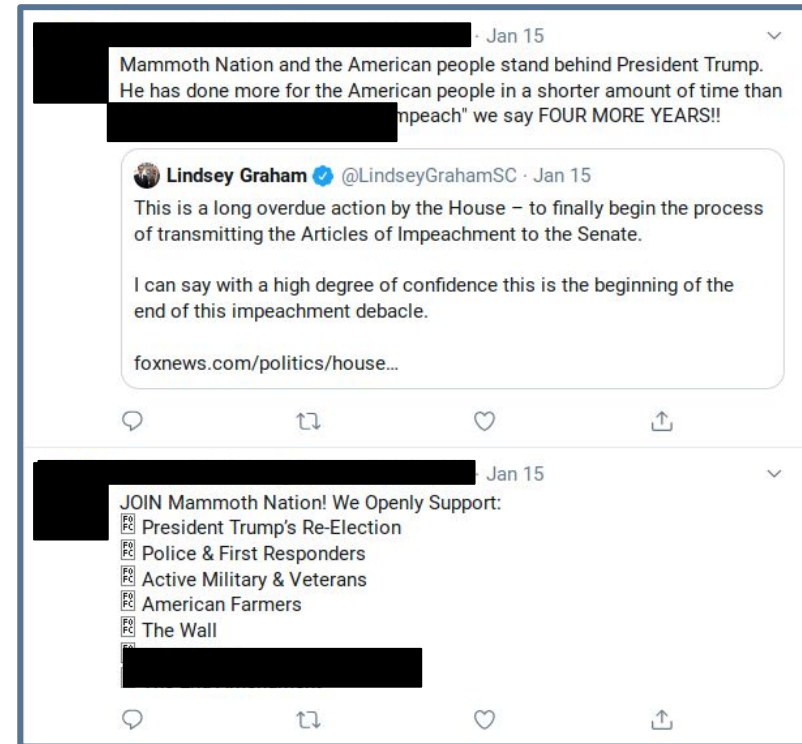
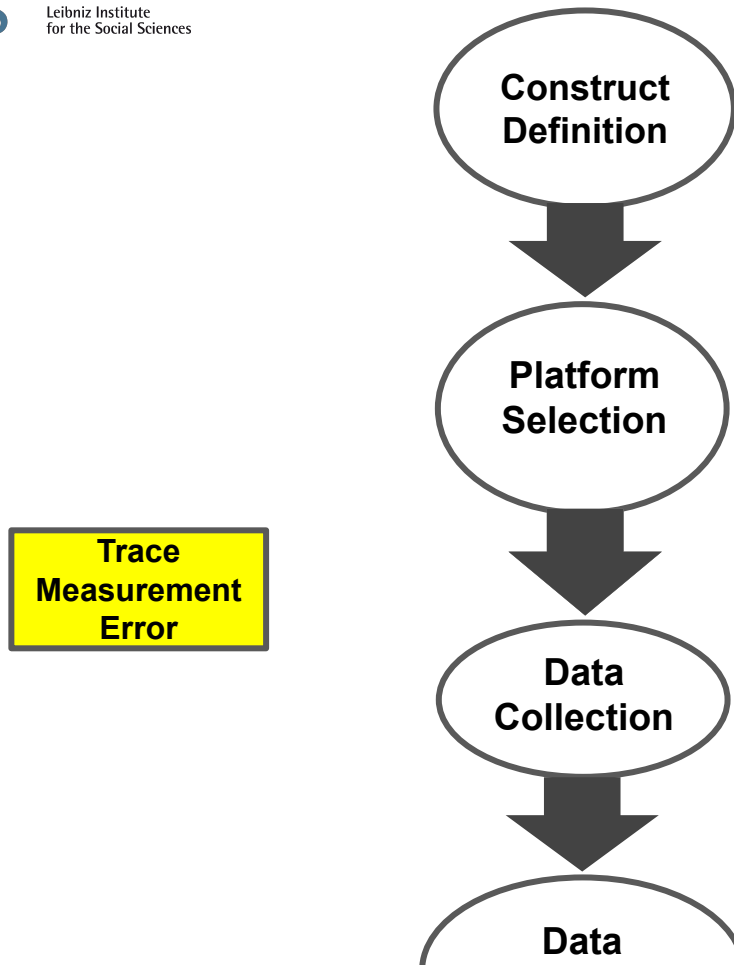
**User  
Reduction  
Error**

**Reduction:** Remove users who are not representative of human users, e.g. bots and organizations

- Karimi, Fariba, et al. "Inferring gender from names on the web: A comparative evaluation of gender detection methods." *Proceedings of the 25th International Conference Companion on World Wide Web* (2016).







- Count the normalized positive words in a day
- The ratio of positive and negative words per tweet
- Add all the ratio of all tweets in a day

Different aggregation methods lead to different measures of approval

## MEASUREMENT

- Ahmed, Saifuddin, Kokil Jaidka, and Marko M. Skoric. "Tweets and votes: A four-country comparison of volumetric and sentiment analysis approaches." *Tenth International AAAI Conference on Web and Social Media* (2016).

### Potential remedies for Trace Measurement Error:

- Assess what different aggregation strategies *mean*
- Consult domain experts

error due to how  
traces are included in  
the model

Trace  
Measurement  
Error

Data  
Analysis







## Who's Tweeting About the President? What Big Survey Data Can Tell Us About Digital Traces?

Josh Pasek<sup>1</sup>, Colleen A. McClain<sup>1</sup>, Frank Newport<sup>2</sup>, and Stephanie Marken<sup>2</sup>

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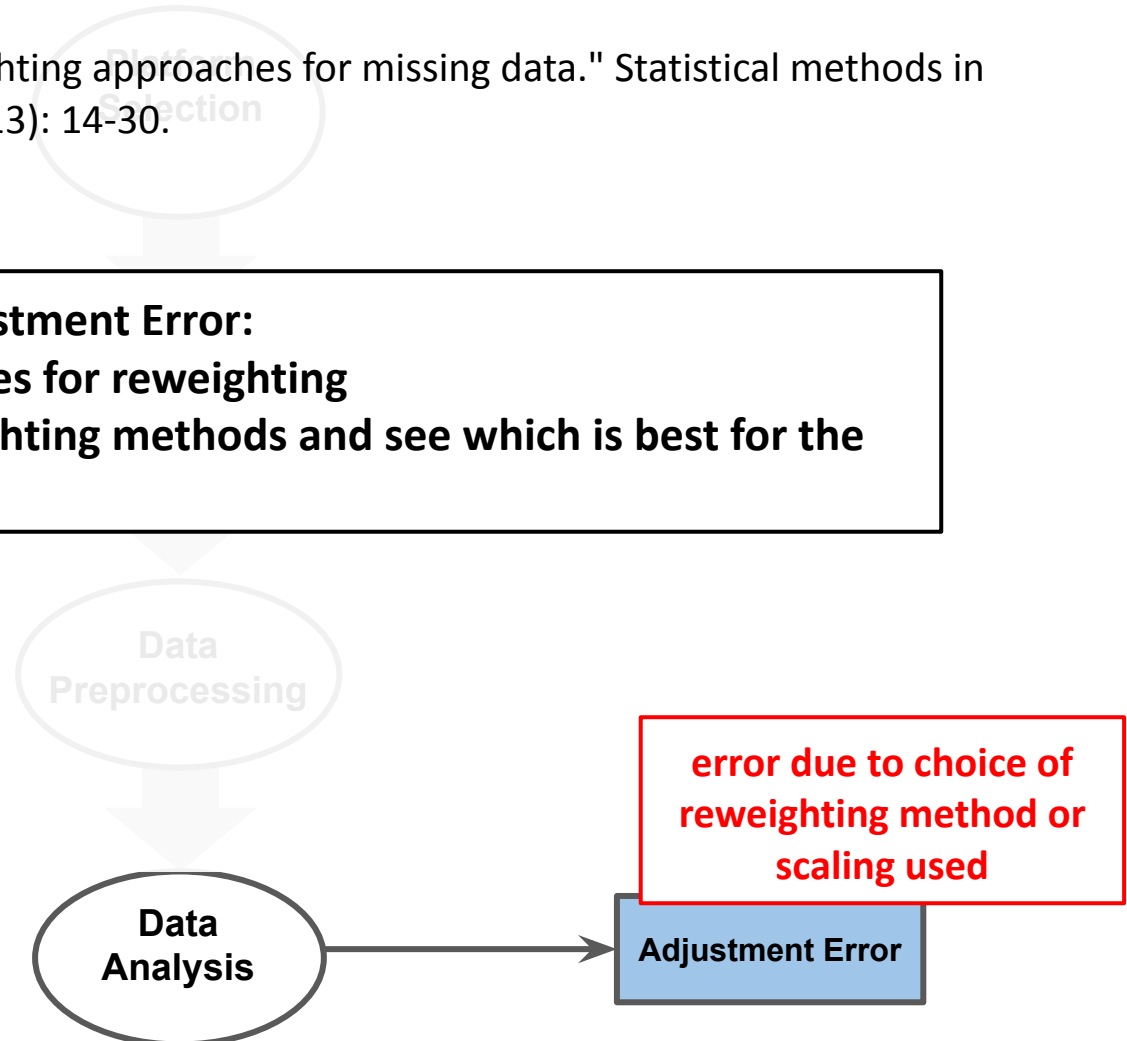
Adjustment  
Error

- Reweighting survey estimates by Twitter demographics to find an alignment may not work since **political Twitterers != random Twitterers**

→ Li, Lingling, et al. "On weighting approaches for missing data." Statistical methods in medical research 22.1 (2013): 14-30.

### Potential remedies for Adjustment Error:

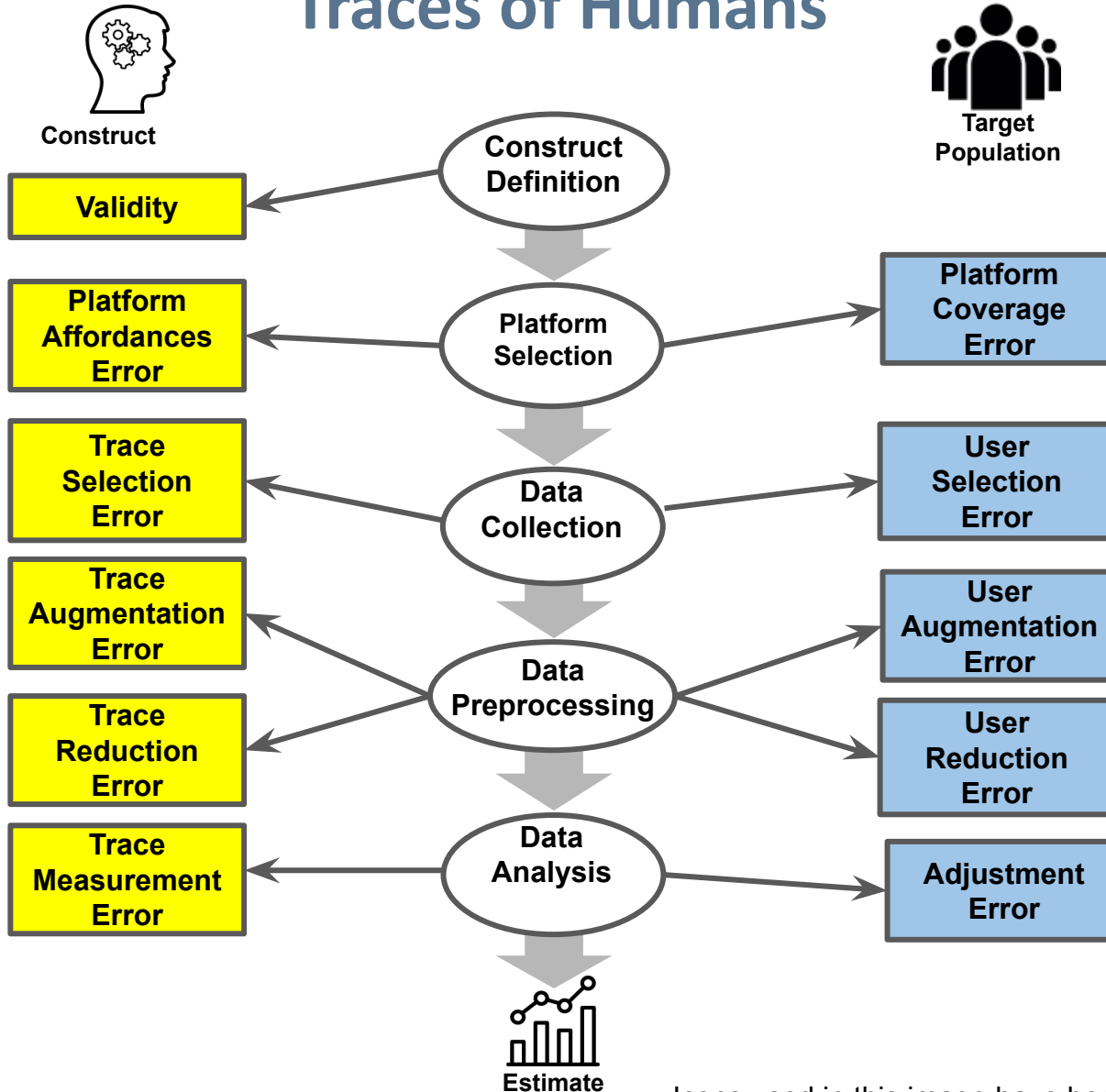
- Carefully select variables for reweighting
- Assess different reweighting methods and see which is best for the problem



# TED-On: A Total Error Framework for Digital Traces of Humans

MEASUREMENT

REPRESENTATION



## To Sum Up...

- TED-On: A Total error framework inspired by survey methodology to decompose errors in DTD-based study
- Acts as guidelines for researchers designing their studies
- A way to systematically illuminate errors - known and unknown, similar to Datasheets [Gebru et al. 2018] and Model Cards [Mitchell et al. 2018]