Detecting Implicitly Harmful Language in Political Discourse

Julia Mendelsohn
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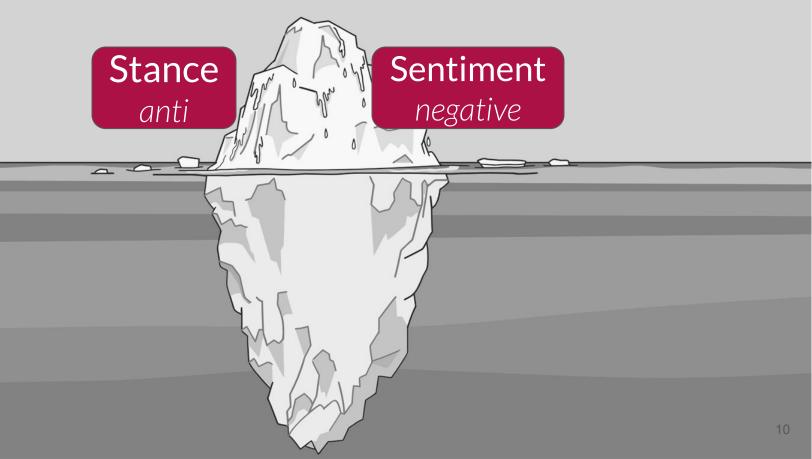
Content Warning

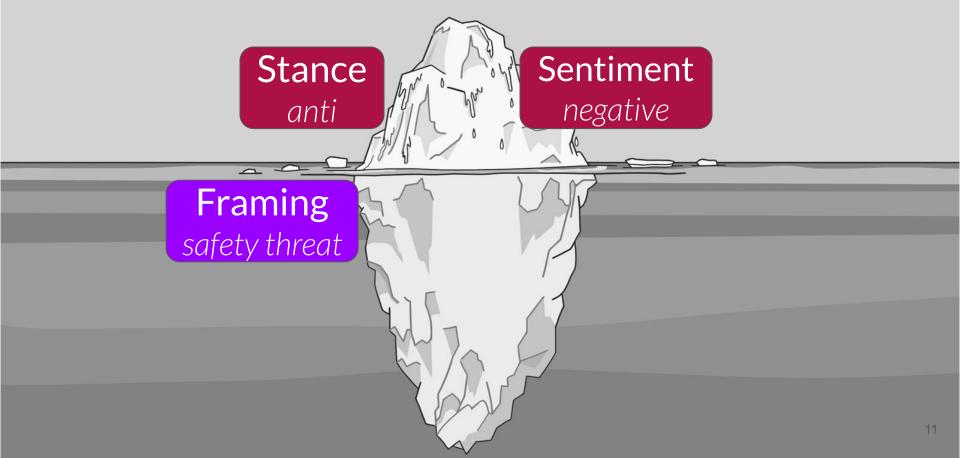
I will be talking about material that may be offensive and upsetting to some audience members.

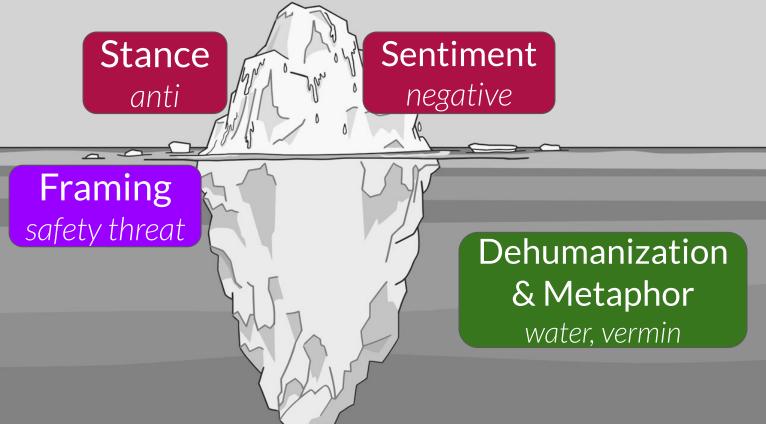


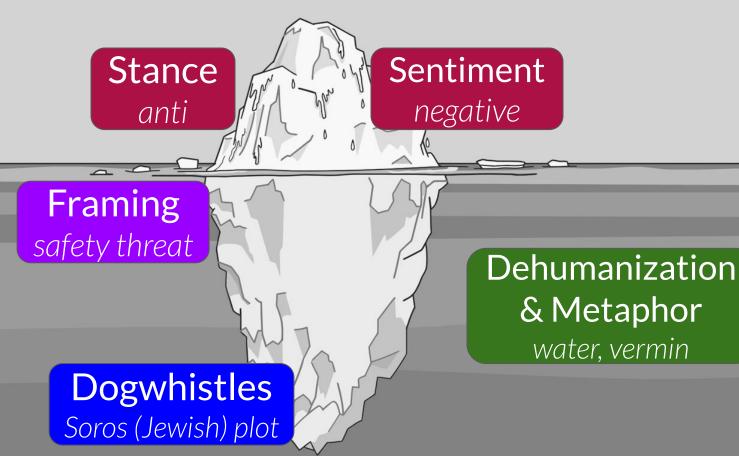


Good night to everyone but the massive amount of Soros-funded i **Anti-immigration**: our border. ~ AZ State Sen Wendy Rogers, Sep 2021

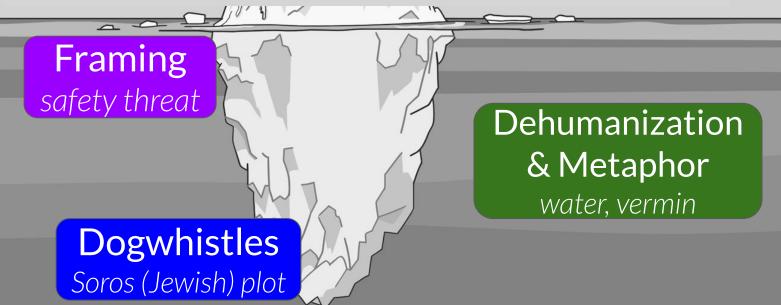


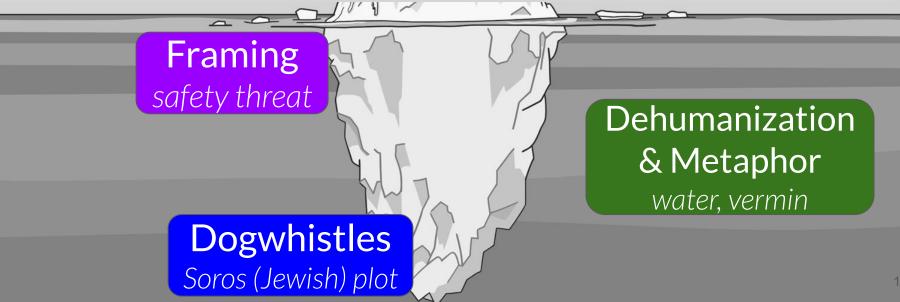






These strategies shape how audiences understand political issues...







Campaigns [Tilley, 2020]





Campaigns [Tilley, 2020]

Media Bias [Esses et al., 2013]



Campaigns [Tilley, 2020]



Media Bias [Esses et al., 2013]



Misinformation
[Henderson &
McCready, 2019]



Campaigns [Tilley, 2020]



Media Bias [Esses et al., 2013]



Misinformation
[Henderson &
McCready, 2019]



Propaganda [Landry et al., 2022]



Electoral Outcomes [Haney López, 2014]



Policymaking [Walgrave et al., 2018]

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Electoral Outcomes

[Haney López, 2014]

Policymaking [Walgrave et al., 2018]



[Jacoby, 2000; Chong & Druckman, 2007]



Policymaking [Walgrave et al., 2018]



Trust [Hopmann et al., 2015]

Electoral Outcomes [Haney López, 2014]



Public
Opinion
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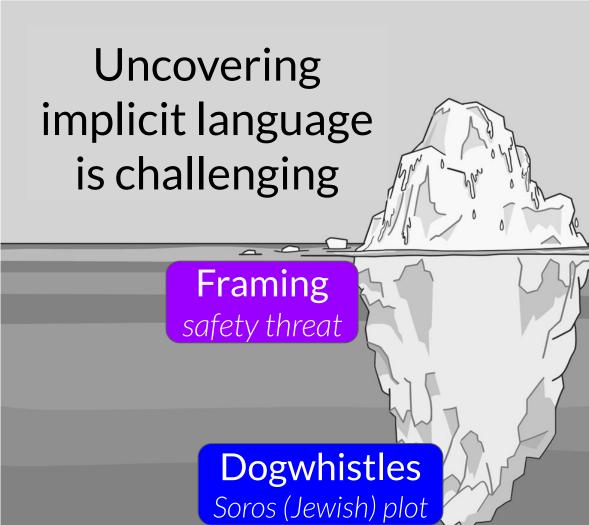
Electoral Outcomes [Haney López, 2014]



Public
Opinion
[Jacoby, 2000; Chong
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Safety & Well-being [Rai et al., 2017]



Dehumanization & Metaphor

water, vermin

Uncovering implicit language is challenging

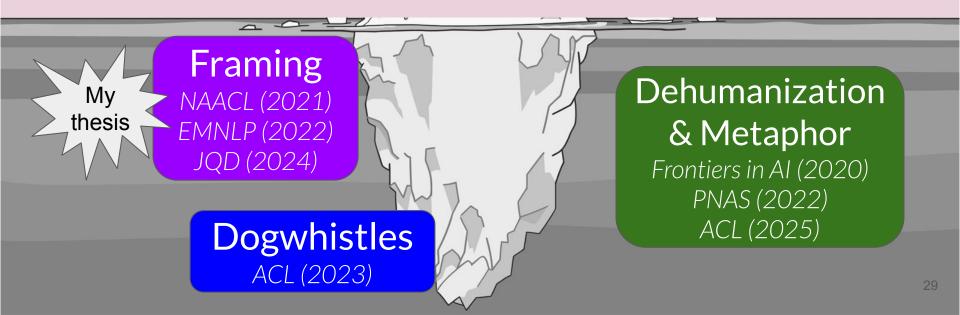
But computational methods can help!

Framing safety threat

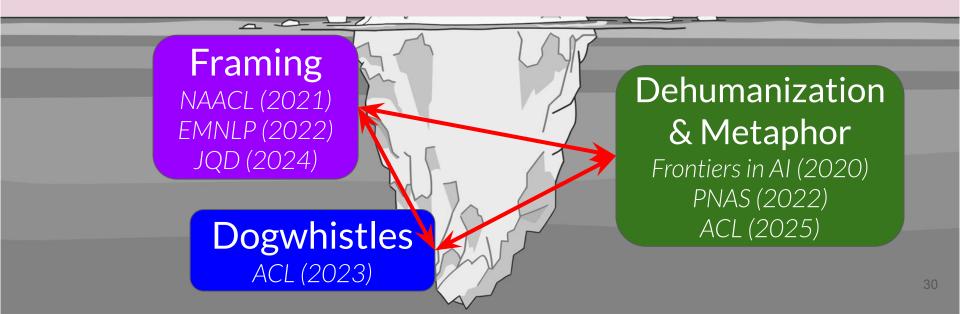
Dogwhistles
Soros (Jewish) plot

Dehumanization & Metaphor water, vermin

I develop computational approaches to study these strategies and their social, political & technological implications



I develop computational approaches to study strategies of othering and their social, political & technological implications



Many ways to think about framing



Modeling Framing in Immigration Discourse on Social Media

Julia Mendelsohn University of Michigan juliame@umich.edu Ceren Budak University of Michigan cbudak@umich.edu David Jurgens

University of Michigan jurgens@umich.edu

Abstract

The framing of political issues can influence policy and public opinion. Even though the public plays a key role in creating and spreading frames, little is known about how ordinary people on social media frame political issues. By creating a new dataset of immigrationrelated tweets labeled for multiple framing typologies from political communication theory, we develop supervised models to detect frames. We demonstrate how users' ideology and region impact framing choices, and how a message's framing influences audience responses. We find that the more commonlyused issue-generic frames obscure important ideological and regional patterns that are only revealed by immigration-specific frames. Furthermore, frames oriented towards human interests, culture, and politics are associated with higher user engagement. This large-scale analysis of a complex social and linguistic phenomenon contributes to both NLP and social science research.

social media content enables us to compare framing strategies across countries and political ideologies. Furthermore, social media provides unique insights into how messages resonate with audiences through interactive signals such as retweets and favorites. By jointly analyzing the production and reception of frames on Twitter, we provide an in-depth analysis of immigration framing by and on the public.

Political communications research has identified numerous typologies of frames, such as issue-generic policy, immigration-specific, and narrative. Each of these frame types can significantly shape the audience's perceptions of an issue (tyengar, 1991; Chong and Druckman, 2007; Lecheler et al., 2015), but prior NLP work seeking to detect frames in mass media (e.g. Card et al., 2016; Field et al., 2018; Kwak et al., 2020) has largely been limited to a single issue-generic policy typology. Multiple dimensions of framing must be considered in order to better understand the structure of immigration discourse and its effect on public opinion

nal of Quantitative Description: Digital Media 4(2024), 1–61 10.51685/jqd.2024.icv

Framing Social Movements on Social Media:
Unpacking Diagnostic, Prognostic, and Motivational Strategies

JULIA MENDELSOHN MAYA VIJAN DALLAS CARD CEREN BUDAK

University of Michigan, USA

Social media enables activists to directly communicate with the public and provides a space for movement leaders, participants, bystanders, and opponents to collectively construct and contest narratives. Focusing on Twitter messages from social movements surrounding three issues in 2018-2019 (guns, immigration, and LGBTQ rights), we create a codebook, annotated dataset, and computational models to detect diagnostic (problem identification and attribution), prognostic (proposed solutions and tactics), and motivational (calls to action) framing strategies. We conduct an in-depth unsupervised linguistic analysis of each framing strategy, and uncover cross-movement similarities in

When People are Floods: Analyzing Dehumanizing Metaphors in Immigration Discourse with Large Language Models

Warning: this paper contains examples of upsetting and offensive content.

Julia Mendelsohn University of Maryland juliame@umd.edu

Abstract

Metaphor, discussing one concept in terms of another, is abundant in politics and can shape how people understand important issues. We develop a computational approach to measure metaphorical language, focusing on immigration discourse on social media. Grounded in qualitative social science research, we identify seven source domain concepts evoked in immigration discourse (e.g. WATER or VERMIN). We propose and evaluate a novel technique that leverages both word-level and document-level signals to measure metaphor with respect to these source domains. We then study the relationship between metaphor, political ideology, and user engagement in 400K US tweets about immigration. While conservatives tend to use dehumanizing metaphors more than liberals, this effect varies widely across source domains. Moreover, creature-related metaphor is associated with more retweets, especially for liberal authors. Our work highlights the potential for computational methods to complement quali-

Ceren Budak University of Michigan chudak@umi.ch..edu



Figure 1: Dehumanizing sentence likening immigrants to the *source domain concepts* of WATER and VERMIN via the words "pour" and "infest".

for measuring and analyzing metaphor at scale. We use this methodology to study dehumanizing metaphor in immigration discourse on social media, and analyze the relationship between metaphor use, political ideology, and user engagement.

From prior discourse analysis literature, we first identify seven source domains: concepts evoked in discussions of immigration such as WATER or VERMIN (Figure 1). We use large language models

Modeling Framing in Immigration Discourse on Social Media

NAACL 2021

Julia Mendelsohn

Ceren Budak

David Jurgens



What is framing?

"Selecting some aspects of a perceived reality and make them more salient in a communicating text, in such a way as to promote a particular problem definition, causal interpretation, moral evaluation, and/or treatment recommendation for the item described" [Entman, 1993]

What is a frame?

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- Issue-generic Policy [Boydstun et al., 2013]
 - Crime & punishment, morality, economic, policy

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- Immigration-specific [Benson, 2013]
 - Immigrants as victims (e.g. of global economy or discrimination)
 - Immigrants as heroes (e.g. contributing to economy or cultural diversity)
 - Immigrants as threats (e.g. to jobs, or to public safety)

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 - Immigrants as threats (e.g. to jobs, or to public safety)
- Issue-generic Narrative [lyengar, 1991]
 - Episodic: focus on specific actions, events, examples, or case studies
 - Thematic: focus on broader political, social, cultural context

Framing processes

Frame-building: factors affecting how an issue is framed



Figure & theoretical model adapted from de Vreese [2005] and is a simplification of Scheufele's [1999] four-process model

Framing processes

- Frame-building: factors affecting how an issue is framed
- Frame-setting: frame effects on audiences

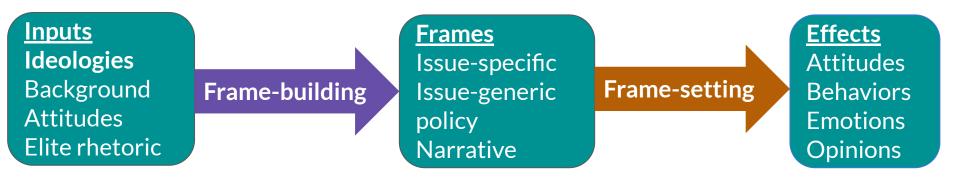


Figure & theoretical model adapted from de Vreese [2005] and is a simplification of Scheufele's [1999] four-process model









Dataset collection & annotation

frame detection

Automated Frame building: role of ideology in framing

Frame setting: effects on user engagement

Data Annotation

3 typologies

27 categories

Frame Type	Frame	Description
Issue-Generic	Economic	Financial implications of an issue
Policy	Capacity & Resources	The availability or lack of time, physical, human, or financial resources
	Morality & Ethics	Perspectives compelled by religion or secular sense of ethics or social responsibility
	Fairness & Equality	The (in)equality with which laws, punishments, rewards, resources are distributed
	Legality, Constitutionality	Court cases and existing laws that regulate policies; constitutional interpretation;
	& Jurisdiction	legal processes such as seeking asylum or obtaining citizenship; jurisdiction
	Crime & Punishment	The violation of policies in practice and the consequences of those violations
	Security & Defense	Any threat to a person, group, or nation and defenses taken to avoid that threat
	Health & Safety	Health and safety outcomes of a policy issue, discussions of health care
	Quality of Life	Effects on people's wealth, mobility, daily routines, community life, happiness, etc.
	Cultural Identity	Social norms, trends, values, and customs; integration/assimilation efforts
	Public Sentiment	General social attitudes, protests, polling, interest groups, public passage of laws
	Political Factors &	Focus on politicians, political parties, governing bodies, political campaigns
	Implications	and debates; discussions of elections and voting
	Policy Prescription &	Discussions of existing or proposed policies and their effectiveness
	Evaluation	
	External Regulation &	Relations between nations or states/provinces; agreements between governments;
	Reputation	perceptions of one nation/state by another
Immigration Specific	Victim: Global Economy	Immigrants are victims of global poverty, underdevelopment and inequality
	Victim: Humanitarian	Immigrants experience economic, social, and political suffering and hardships
	Victim: War	Focus on war and violent conflict as reason for immigration
	Victim: Discrimination	Immigrants are victims of racism, xenophobia, and religion-based discrimination
	Hero: Cultural Diversity	Highlights positive aspects of differences that immigrants bring to society
	Hero: Integration	Immigrants successfully adapt and fit into their host society
	Hero: Worker	Immigrants contribute to economic prosperity and are an important source of labor
	Threat: Jobs	Immigrants take nonimmigrants' jobs or lower their wages
	Threat: Public Order	Immigrants threaten public safety by being breaking the law or spreading disease
	Threat: Fiscal	Immigrants abuse social service programs and are a burden on resources
	Threat: National Cohesion	Immigrants' cultural differences are a threat to national unity and social harmony
Narrative	Episodic	Message provides concrete information about on specific people, places, or events
	Thematic	Message is more abstract, placing stories in broader political and social contexts

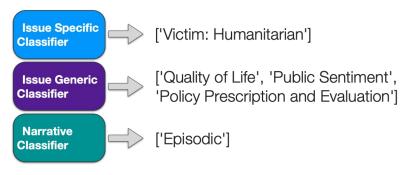
Data Collection and Annotation

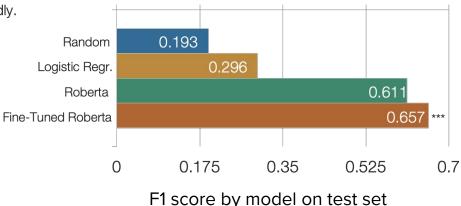
- 2.6M English tweets from 2018-2019 containing immigration-related term
- Ideology inference using existing network-based tool
- 4.5K tweets labeled by trained annotators for all frames explicitly cued.

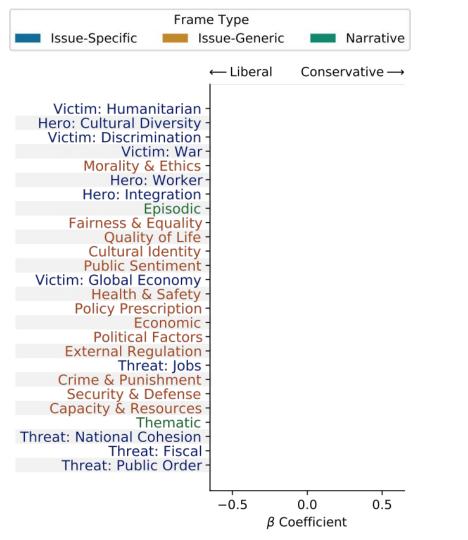
Frame Detection

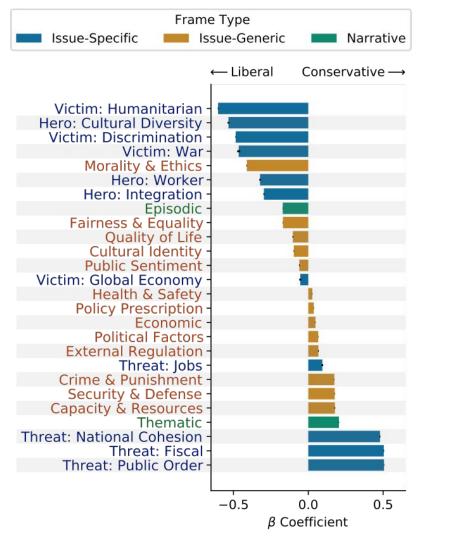
 RoBERTa-based multilabel classification models fine-tuned on full set of immigration-related tweets

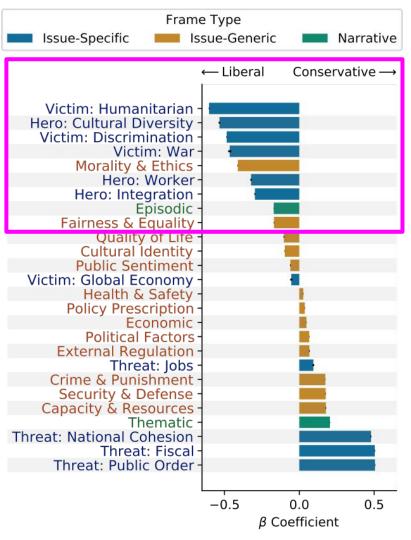
"The proposed #TitleIX rule will exacerbate the negative experiences of undocumented students on campus. Get ready to speak your voice loudly. Go to https://t.co/7kAlhYjeLl to learn how with @endrapeoncampus"





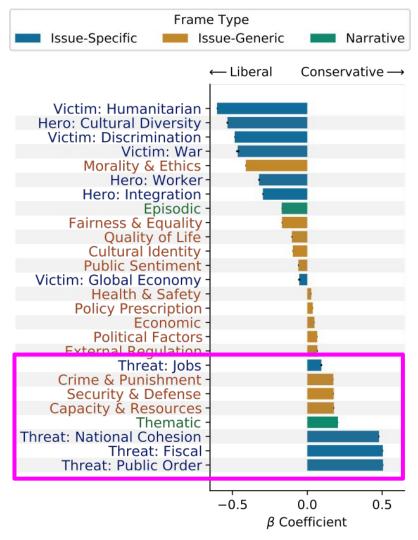






Liberals frame immigrants as heroes and victims

 Liberals cue fairness and morality, framing immigrants as victims of discrimination and inhumane policies.

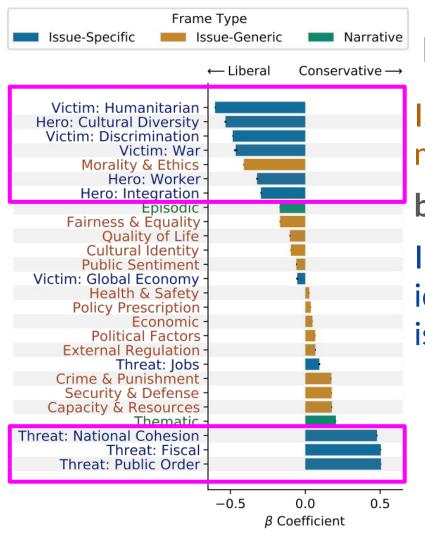


Liberals frame immigrants as heroes and victims

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Conservatives frame immigrants as **threats**

 Conservatives cue threat to public safety, burden on taxpayers & government programs

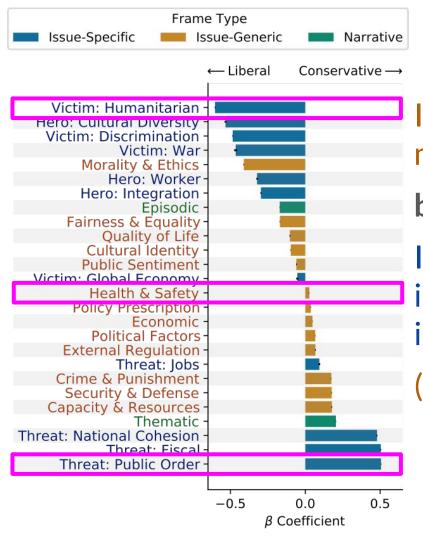


Each frame typology offers value

Issue-generic policy frames can be most accurately detected

but

Immigration-specific frames reveal ideological differences obscured by issue-generic policy frames



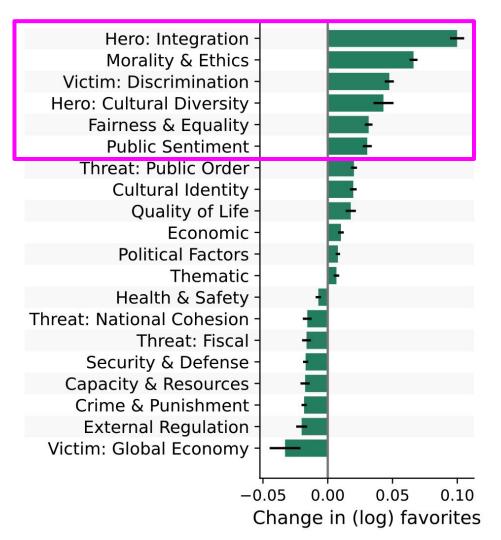
Each frame typology offers value

Issue-generic policy frames can be most accurately detected

but

Immigration-specific frames reveal ideological differences obscured by issue-generic policy frames

(e.g. health & safety)





Cultural (hero: integration) and human interest (morality, fairness, victim: discrimination)

Framing Social Movements on Social Media: Unpacking Diagnostic, Prognostic, and Motivational Strategies

Journal of Quantitative Description: Digital Media (2024)

Julia Mendelsohn Dallas Card Maya Vijan Ceren Budak





Adbusters Magazine @Adbusters



September 17th. Wall Street. Bring Tent. http://bit.ly/re9ENL

#OCCUPYWALLSTREET

Reply 13 Retweet * Favorite







If you've been sexually harassed or assaulted write 'me too' as a reply to this tweet.

Me too.

Suggested by a friend: "If all the women who have been sexually harassed or assaulted wrote 'Me too.' as a status, we might give people a sense of the magnitude of the problem."

1:21 PM - 15 Oct 2017





marchforourlivesaz.org

TEXT 'MARCH' TO 90975

@PHXMai



Who's wearing red tomorrow? Show your support for marriage equality -make your profile image red for tomorrow and check out www.hrc.org/StandForMarriage for more ways to get involved!



Social movements create meaning through framing

- Collective action frames are "intended to mobilize potential adherents and constituents, to garner bystander support, and to demobilize antagonists" [Snow & Benford, 1988]
- Effective framing is important (perhaps, necessary) for social movement success [Della Porta & Diani, 2006]

Core Framing Tasks [Snow & Benford, 1988; Benford & Snow, 2000]

Diagnostic

Identifying social problems, their causes, and who to hold responsible

Core Framing Tasks [Snow & Benford, 1988; Benford & Snow, 2000]

Diagnostic

Identifying social problems, their causes, and who to hold responsible

Prognostic

Proposing solutions, plans of attack, strategies for carrying out that plan

Core Framing Tasks [Snow & Benford, 1988; Benford & Snow, 2000]

Diagnostic

Identifying social problems, their causes, and who to hold responsible

Prognostic

Proposing solutions, plans of attack, strategies for carrying out that plan

Motivational

Persuading people to participate through "calls to action"

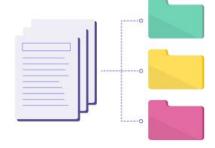
How do people use diagnostic, prognostic, and motivational framing in Twitter messages related to social movements?



Dataset & Annotation



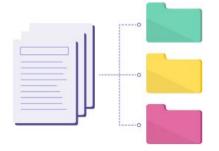
Dataset & Annotation



Classification



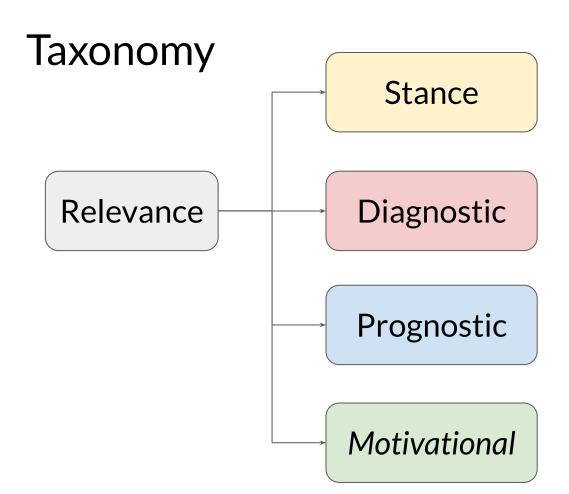
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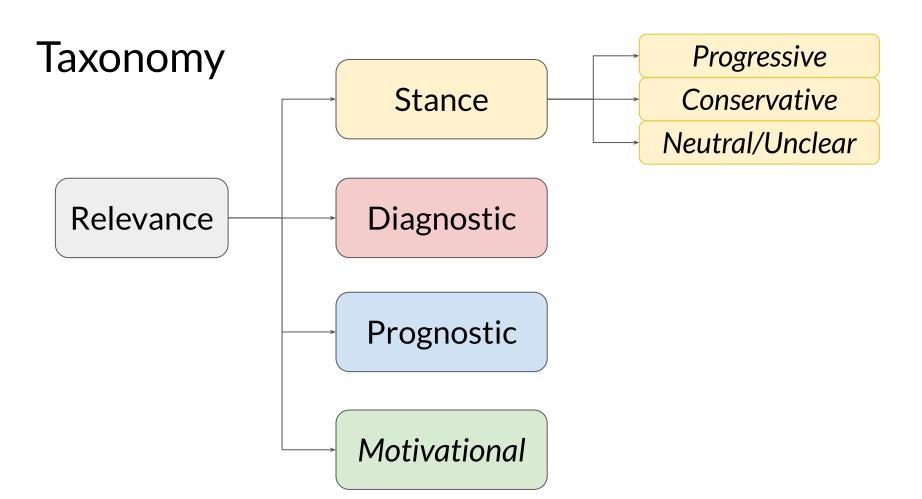


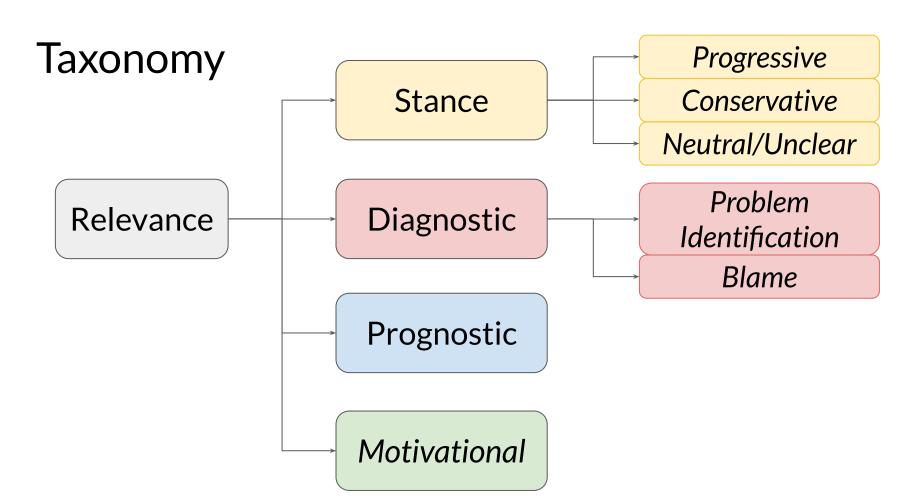
Classification

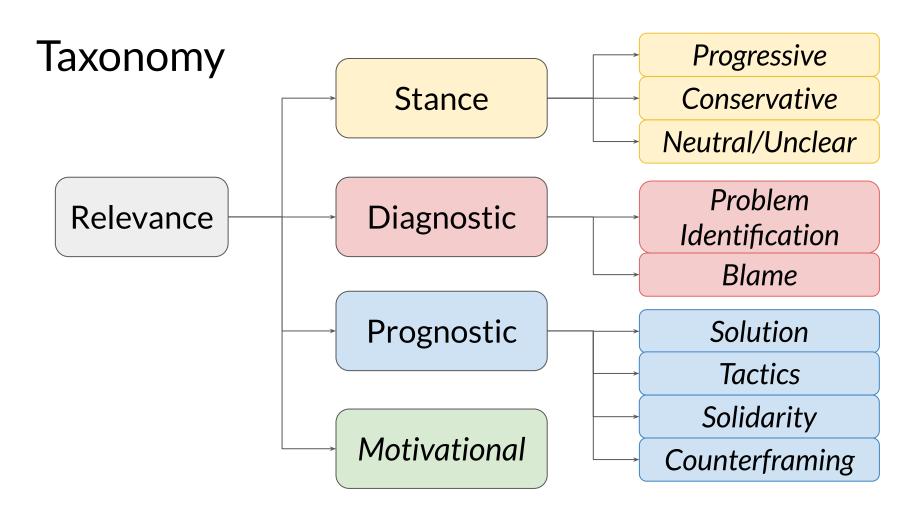


Sociocultural Context









Data Collection [from Bozarth & Budak, 2022]

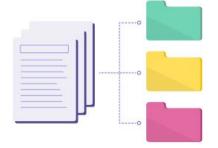
- Tweets from movements focused on 3 issues: guns, immigration, and LGBTQ rights from 2018-2019
 - Both progressive & conservative movements
 - 2 months for each issue (1 high protest activity, 1 average level)

Data Collection [from Bozarth & Budak, 2022]

- Tweets from movements focused on 3 issues: guns, immigration, and LGBTQ rights from 2018-2019
 - Both progressive & conservative movements
 - 2 months for each issue (1 high protest activity, 1 average level)
- 1.85M tweets across all movements
 - 822K for guns, 763K for immigration, 268K for LGBTQ
- 6,000 manually-annotated tweets
- 4,859 (81%) coded as *relevant*, labeled for stance & frames



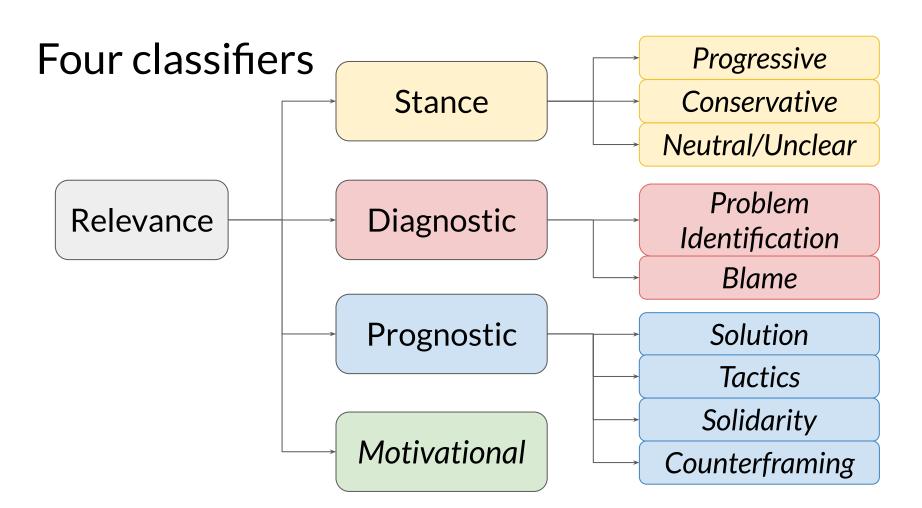
Dataset & Annotation



Classification



Sociocultural Context

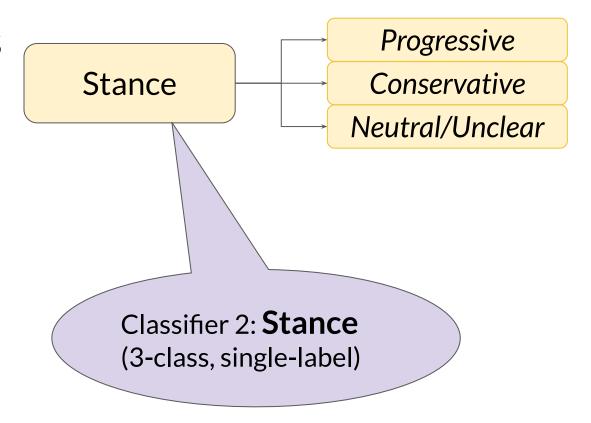


Four classifiers

Relevance

Classifier 1: **Relevance**(binary, single-label)

Four classifiers



Four classifiers Classifier 3: **Core Framing Task** (binary, 3-label) Diagnostic Prognostic **Motivational**

Four classifiers

Classifier 4:

Frame Elements

Categories we coded for (binary, 7-label)

Motivational

Problem Identification

Blame

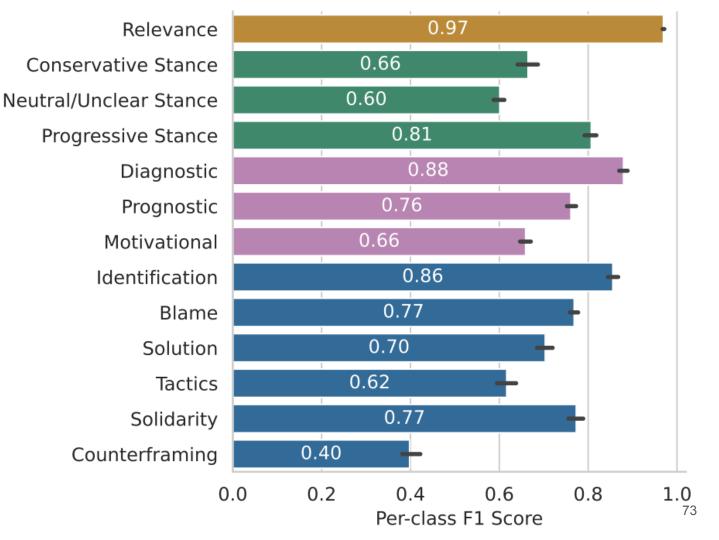
Solution

Tactics

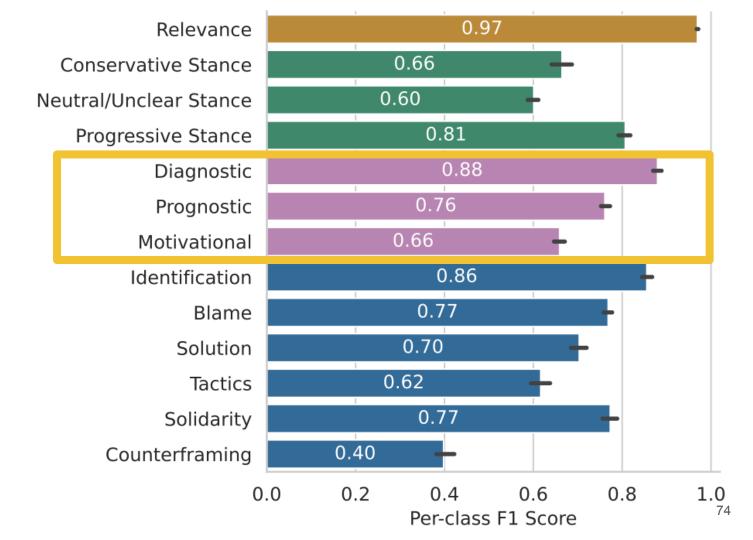
Solidarity

Counterframing

Model Results



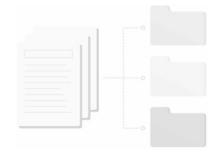
Model Results





Dataset &

Annotation



Classification



Sociocultural Context

3 movement-level factors

2 message-level factors

Issue Area guns, immigration, LGBTQ

Stance progressive, conservative, neutral

Protest Activity

High or average activity month

Author Role journalist, social mvmt org, other

Interaction Type broadcast, reply, quote tweet

Logistic regression + marginal effect for diagnostic, prognostic, and motivational frames

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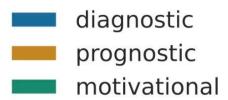
Protest Activity

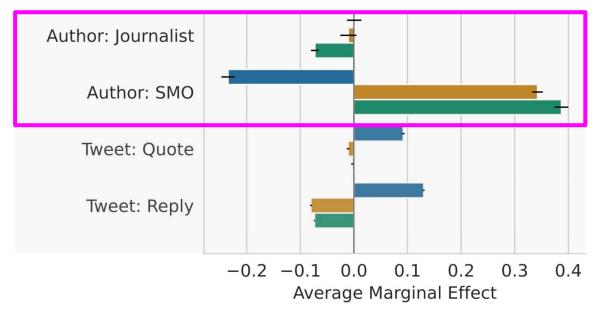
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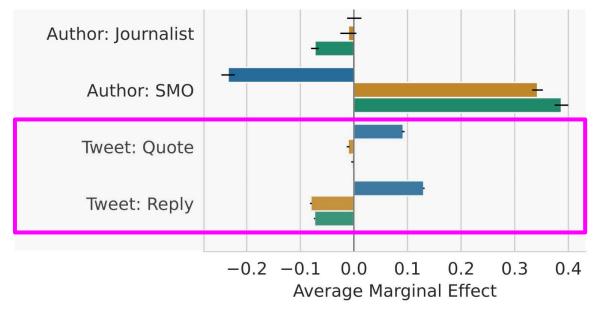
Diagnostic: not SMOs

Prognostic: SMOs

Motivational: SMOs

"Other" is reference variable

diagnosticprognosticmotivational



Diagnostic: QT & reply

Prognostic: QT &

broadcast

Motivational: QT &

broadcast

^{*}broadcast is reference variable

Framing Social Movements on Social Media:

Unpacking Diagnostic, Prognostic, and Motivational Strategies



Not discussed today: linguistic analysis within each frame category

I spent *many months* manually annotating immigration-related tweets (over 10K tweets across papers!) and saw some really weird stuff....



I saw tons of tweets covertly blaming Jews for the immigration "crisis", but nobody seemed to notice

From Dogwhistles to Bullhorns: Unveiling Coded Rhetoric with Language Models

Association for Computational Linguistics (ACL), 2023



Julia Mendelsohn



Ronan Le Bras



Yejin Choi



Maarten Sap

The cosmopolitan elite look down on the common affections that once bound this nation together: things like place and national feeling and religious faith...The cosmopolitan agenda has driven both Left and Right...It's time we ended the cosmopolitan experiment and recovered the promise of the republic. ~Josh Hawley (R-MO), 2019



The **Jews** look down on the common affections that once bound this nation together: things like place and national feeling and religious faith...The **Jewish** agenda has driven both Left and Right...It's time we ended the **Jewish** experiment and recovered the promise of the republic. ~*Josh Hawley (R-MO), 2019*



Cosmopolitan is a dogwhistle

Dogwhistles send one message to an outgroup and a second (often taboo, controversial, or inflammatory) message to an in-group [Henderson & McCready, 2018]

Cosmopolitan is a dogwhistle

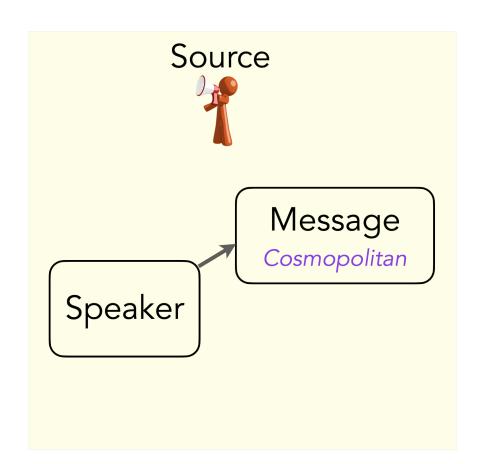
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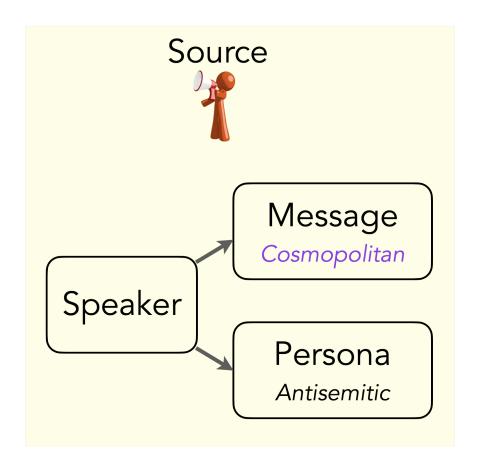
In-group knows cosmopolitan → Jewish

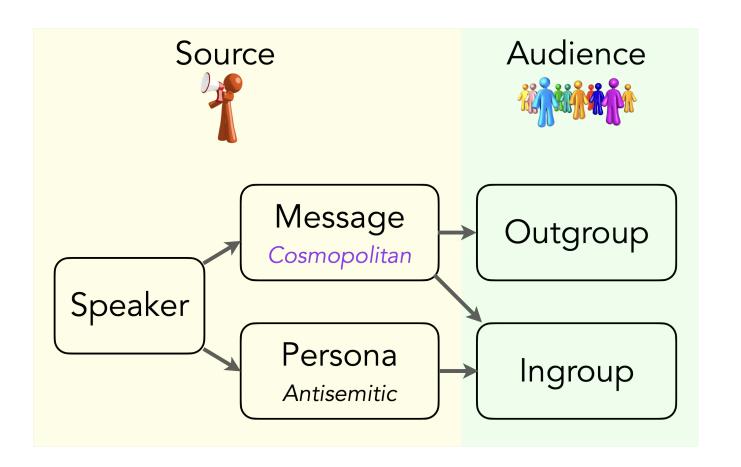
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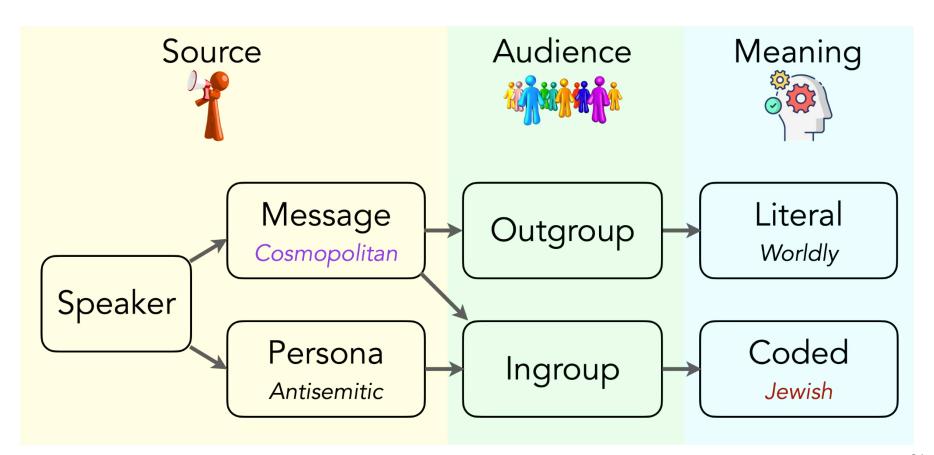
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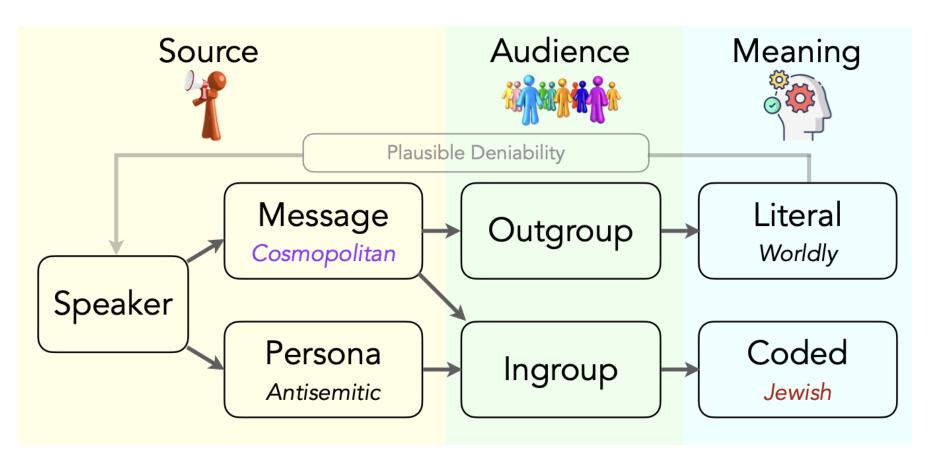
- In-group knows cosmopolitan → Jewish
- But Hawley has plausible deniability. He never says Jewish!

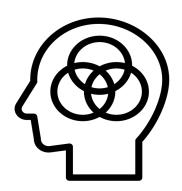






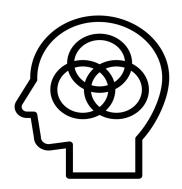






Meaning depends on speaker identity, context, and multiple audiences

94

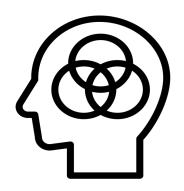


Meaning depends on speaker identity, context, and multiple audiences 0

Mechanism of political influence and persuasion

[Mendelberg, 2001; Haney López, 2014]

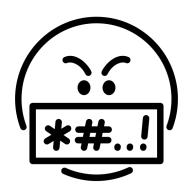
[Henderson & McCready, 2018]



Meaning depends on speaker identity, context, and *multiple* audiences 0

Mechanism of political influence and persuasion

[Mendelberg, 2001; Haney López, 2014]



Enables hate while evading content moderation

[Bhat & Klein, 2020]







Evaluate
dogwhistle
recognition in
language models





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Show how dogwhistles evade content moderation





Evaluate dogwhistle recognition in language models



Show how dogwhistles evade content moderation

Searching for dogwhistles

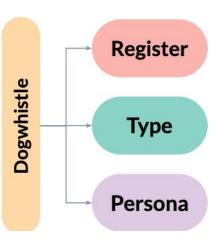
- Sources: academic, media, blogs, wikis
 - Expressions identified as dogwhistles or coded language

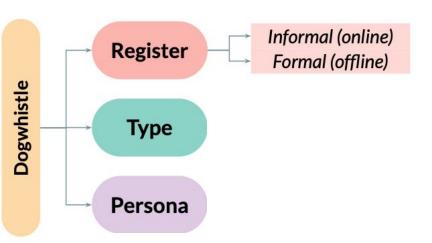
Searching for dogwhistles

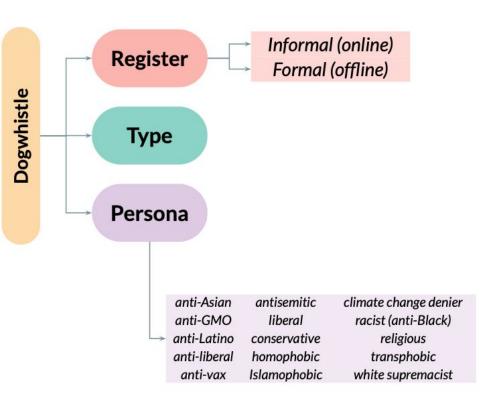
- Sources: academic, media, blogs, wikis
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- 340 terms and symbols (incl. emojis)
 - Over 70 each for racist, transphobic, antisemitic
 - English, US-centric

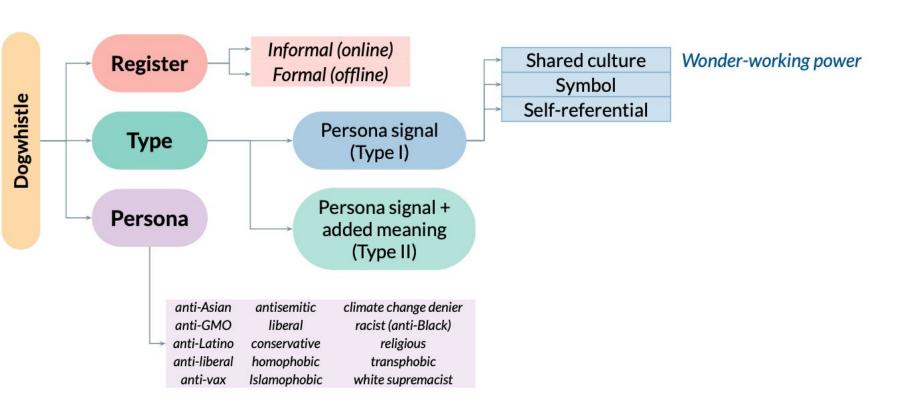
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- Limitation: we cannot ensure that our search is complete or figure out what's missing.
 - Can large language models help? Stay tuned...

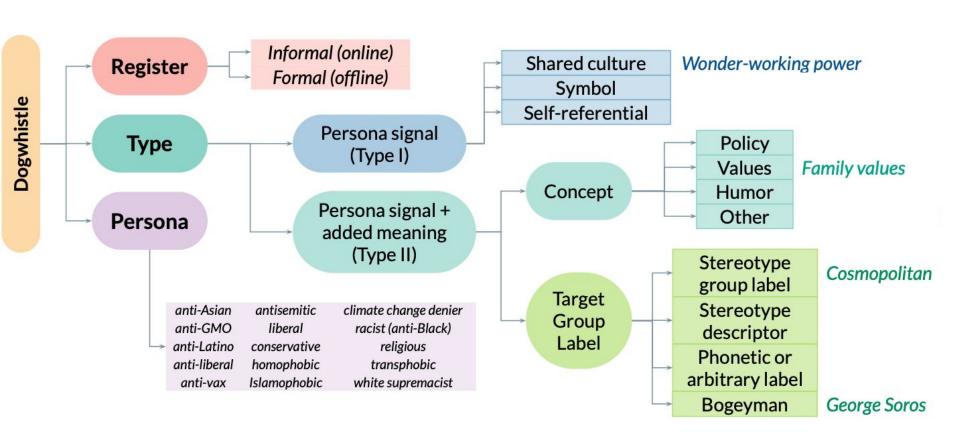








^{*}Type I and Type II distinction from Henderson & McCready (2018)



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Sex-based rights
Trans people threaten cis women's rights
Transphobic
Concept: Value
Formal

Dogwhistle	Sex-based rights		
In-group meaning	Trans people threaten cis women's rights		
Persona	Transphobic		
Type	Concept: Value		
Register	Formal		
Explanation	Many anti-transgender people [claim that] women's "sex-based rights" are somehow being threatened, removed, weakened, eroded, or erased by transgender rights"Sex-based rights", by the plain English meaning		
Source	Medium post by David Allsopp		

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Source	Medium post by David Allsopp		
Example	When so-called leftists like @lloyd_rm demand that we give up our hard won sex-based rights, they align themselves squarely with men's rights activists. To both groups, female trauma is white noise, an irrelevance, or else exaggerated or invented.		
Context	Tweet by J.K. Rowling on June 28, 2020		



Typology & glossary with rich contextual information



Evaluate
dogwhistle
recognition in
language models



Show how dogwhistles evade content moderation

Our Prompt A dogwhistle is the use of coded or suggestive language in political messaging to garner support from a particular group without provoking opposition. What are examples of dogwhistles?

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1. "Law and order"

- 2. "The silent majority"
- 3. "Family values"
- 4. "Welfare queens"
- 5. "Illegal aliens"

GPT-3 Completion

 Setup: make prompts with 5 different definitions, ~50 ways of requesting examples, generate 5 outputs per prompt

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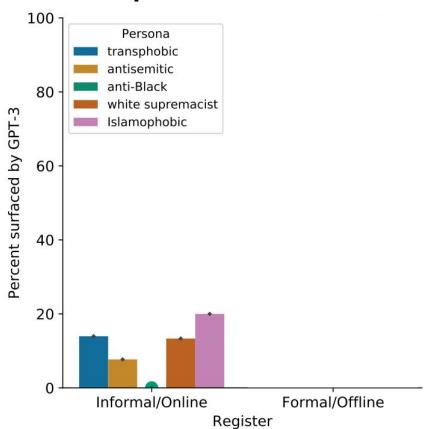
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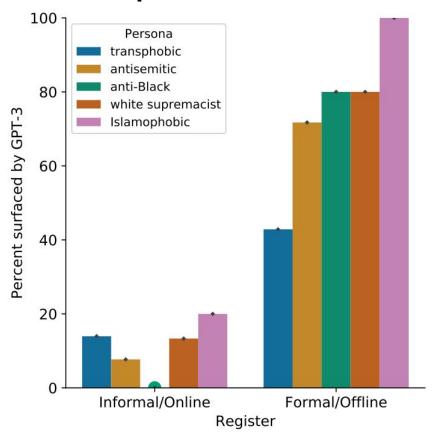
Surfaced additional potential dogwhistles (e.g. patriotism)

But performance varies a lot



 GPT-3 surfaces 69% of formal but <20% of informal dogwhistles

But performance varies a lot



GPT-3 surfaces 69% of formal but <20% of informal dogwhistles

 Among formal dogwhistles, lowest recall for transphobic

Identifying covert meanings with GPT-3

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GPT-3 Completion "Jewish to many anti-Semitic people.

Identifying covert meanings with GPT-3

A dogwhistle is the use of coded or suggestive language in political messaging to garner support from a particular group without provoking opposition. For example, "cosmopolitan" secretly means "Jewish to many anti-Semitic people.

	No Secret Cue	With Secret Cue
No Definition	8.5%	29.6%
With Definition	43.7%	54.3%



Typology & glossary with rich contextual information



Evaluate dogwhistle recognition in language models



Show how dogwhistles evade content moderation

Case Study: Toxicity Detection

Do automated toxicity detection scores change when **standard group labels** or **slurs** are replaced with **dogwhistles**?

Using the Google/Jigsaw Perspective API model

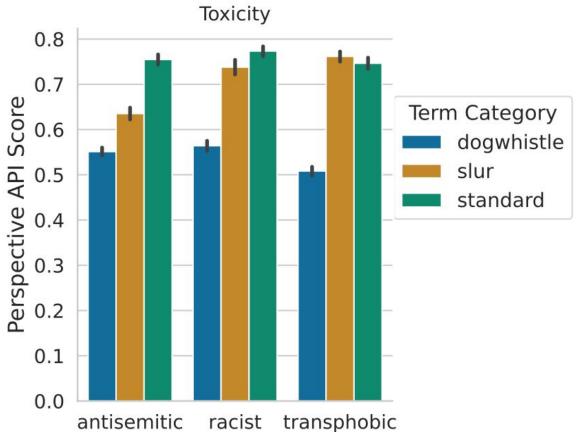
237 hateful templates from HateCheck [Röttger et al., 2021]

Template

So sick of being around [IDENTITY_P] like you. [IDENTITY_P] are disgusting. You [IDENTITY_P] are all dishonest. [IDENTITY_P] are human garbage. Your [IDENTITY_A] lives are worthless. *Never trust a single word said by a [IDENTITY_S].* I will cut their [IDENTITY A] throats. I would love to see those [IDENTITY_P] suffer. All [IDENTITY_P] should fuck off forever. Of course that dumbass was a [IDENTITY_S]. [IDENTITY_P] think they're special. I hate them. *Isn't it typical of a [IDENTITY_S] to be that stupid?*

Racist	Antisemitic	Transphobic
African-American	Jewish	transgender
Black	Jew	trans
inner-city	cosmopolitan	AGP
welfare queen	cosmopolitan elite	autogynephile
thug	coastal elite	AAP
dindu	globalist	autoandrophile
gangbanger	international banker	clownfish
ghetto	cultural Marxist	troon
urban	Hollywood elite	TIF
superpredator	Khazar	TIM
n****r	k**e	t****y
c**n	h**b	s****e

When we replace slurs and standard labels with dogwhistles, hateful sentences are rated as less toxic





Typology & glossary with rich contextual information



Evaluate
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Show how dogwhistles evade content moderation

Not discussed today: a case study of racial dogwhistles in historical U.S. political speeches

This work is starting point for many future directions!

- Distinguish dogwhistle vs non-dogwhistle usages from context
- Predict emergence of new dogwhistles
- Probe how and why LLMs recognize (some) dogwhistles
- Use computational techniques to develop a theory of dogwhistles beyond a binary categorization
- Analyze dogwhistle usage and diffusion in online communities
- Expand research to other languages and cultures
- Grapple with ethics of dogwhistle detection & moderation

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- Step 4: Use models to do interesting and important things
 - But...in today's climate?!?!

Two-pronged systematic review

Computational antisemitism work across disciplines/venues

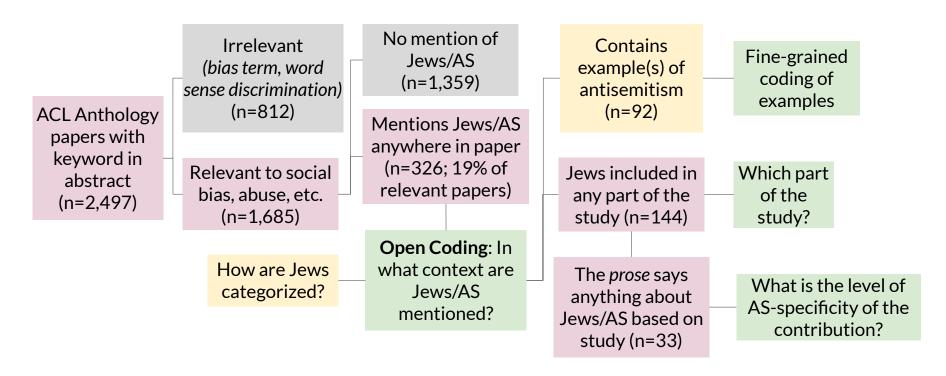
Goal: in-depth description of how antisemitism has been defined, measured, and analyzed

Method: Querying Google Scholar, Scopus, Web of Science NLP work on bias and abusive language mentioning Jews

Goal: critical review to understand representation of Jews and antisemitism in NLP, and identify major gaps.

Method: ACL Anthology search

I read a lot of bias and hate speech papers



Manually coded examples reproduced in NLP papers

Classic antisemitism: tropes that have persisted for centuries

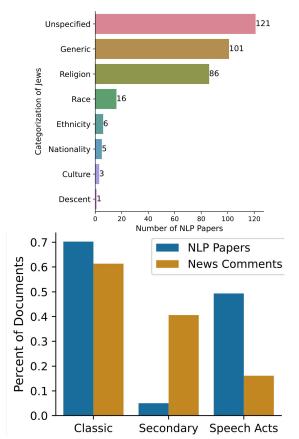
- Examples: foreignness, repulsiveness, evil, greed, power
 Secondary antisemitism: post-Holocaust tropes emerged from guilt, with effects of rejecting Jewish experiences
 - Examples: Holocaust distortion, blaming Jews for antisemitism, denial and instrumentalization of contemporary antisemitism

Aggressive speech acts:

 Examples: insults, death wishes, affirmation of Nazis (e.g. via Holocaust jokes)

Antisemitism is inadequately represented in NLP

- Only 4 papers (0.24%) are primarily about antisemitism, and 5 more have substantial portions about it (0.3%)
- NLP treats Jews as one of many interchangeable groups or solely as religious identity.
- Examples in NLP papers highlight neglect of secondary antisemitism and over-indexing on greed, Holocaust jokes, and insults
- We have no idea how well our models work on "real-world" antisemitism! Or how to evaluate if our models are antisemitic!



Challenges and opportunities for NLP

Socially-grounded bias evaluation

Focus on fine-grained topoi

Understand highly contextual implicit forms (e.g. dogwhistles)

Counter narratives that combat deeper

antisemitic logics

We need resources!

Develop adaptable and explainable models

Ethics-forward design

Home > Newsroom > All News > News Releases > DHS to Begin Screening Aliens' Social Media Activity for Antisemitism

DHS to Begin Screening Aliens' Social Media Activity for Antisemitism

Release Date: 04/09/2025

Community-level recommendations

- Include Jewish people and/or experts, especially in annotation
- We need group-specific approaches to abusive language. Prioritizing generalizability is bad for science and society.
- Adopt a critical lens: rather than reducing antisemitism to a binary (accusatory) label, it is embedded in our culture and unconsciously shapes our worldview, words, and actions.
- Both the dismissal and weaponization of antisemitism are scary. But the contentious political climate is more of a reason to study it.

Thank You!



