

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

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#### Social movements create meaning through framing

- "Schemata of interpretation" that help people "locate, perceive, identify, and label" new information [Goffman, 1974]
- Collective action frames are further "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 blame or hold responsible.
- Prognostic: articulating proposed solutions, plans of attack, and strategies for carrying out that plan
- Motivational: persuading people to participate in the movement through "calls to action"

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

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We develop a codebook, labeled dataset, and supervised learning models to identify framing strategies in tweets.

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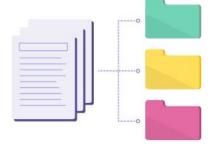
Our analysis unpacks relationships between framing, sociocultural context, and discursive processes



Dataset & Annotation



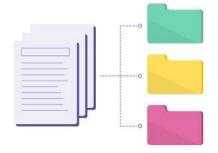
Dataset & Annotation



Classification



Dataset & Annotation



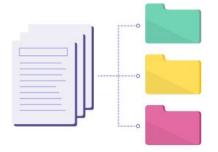
Classification



Analysis: Sociocultural Context



Dataset & Annotation



Classification



Analysis: Sociocultural Context



Analysis: Linguistic Properties



Dataset & Annotation



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Analysis: Linguistic Properties

#### Data Collection [from Bozarth & Budak, 2022]

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

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- Validated keyword expansion with seed keywords from frequent hashtags during collective action events
  - Event information (incl. issue) from Crowd Counting Consortium
- 1.85M tweets across all movements
  - 822K for guns, 763K for immigration, 268K for LGBTQ

# Guns

# Immigration





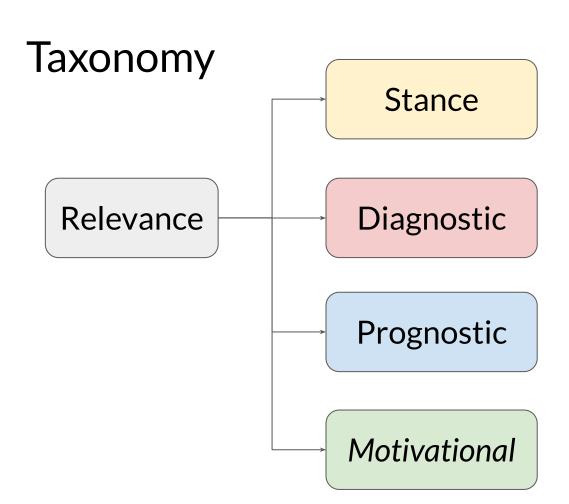


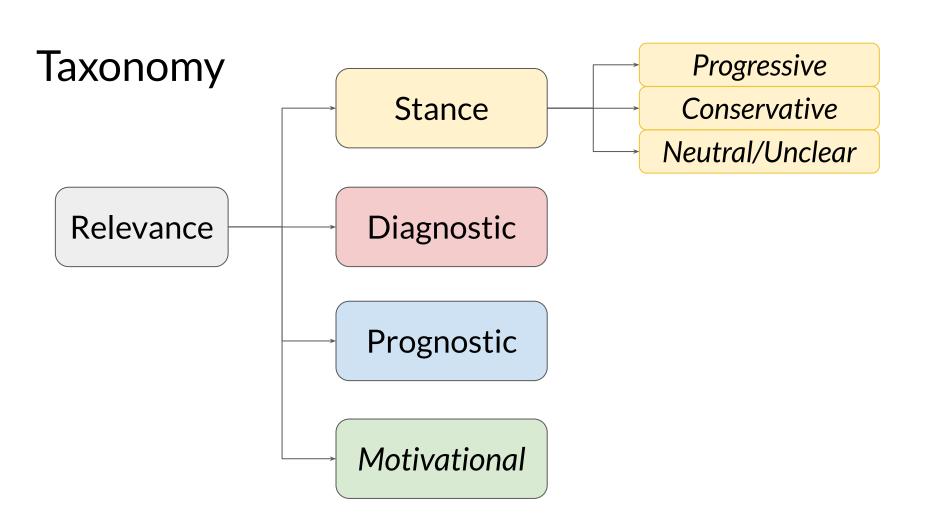


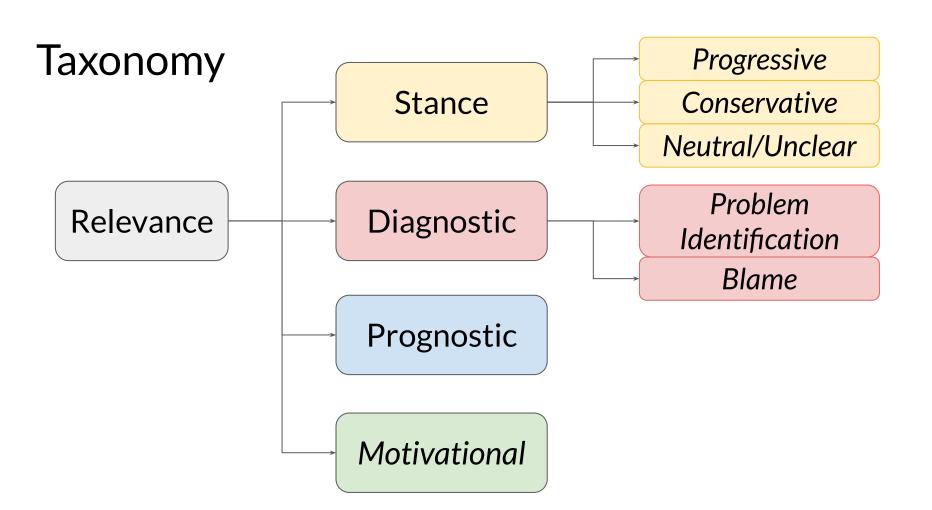


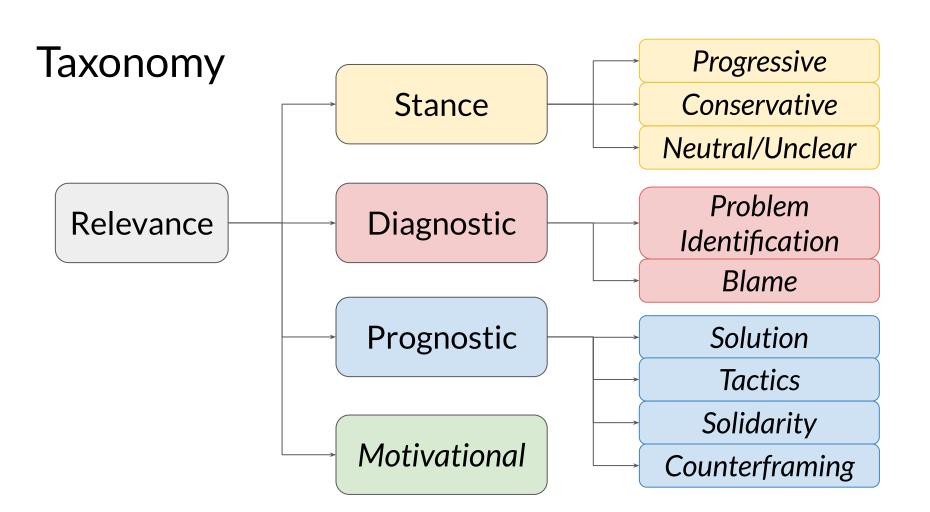








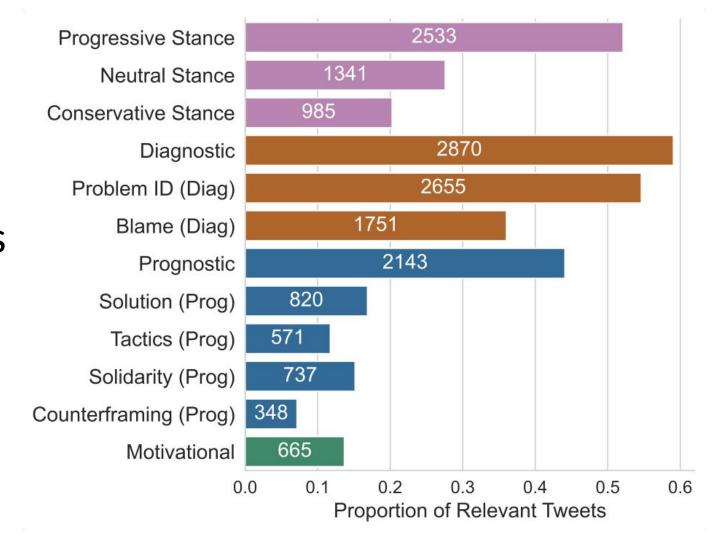




#### **Annotation Procedure**

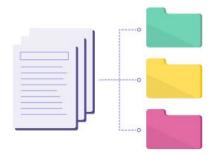
- Preliminary codebook based on theory & initial 150 tweets
- Annotator Training
  - 2 annotators independently labeled 150 tweets, evenly split between issues and activity periods.
  - All disagreements resolved via consensus-coding
  - Sufficient agreement reached after 5 rounds (Krippendorff  $\alpha \ge .7$ )
- Annotators independently coded remaining data
- 6,000 tweets in labeled dataset, evenly split between movements and activity periods
- 4,859 (81%) coded as *relevant*, labeled for stance & frames

Labeled
Data
Statistics





Dataset & Annotation



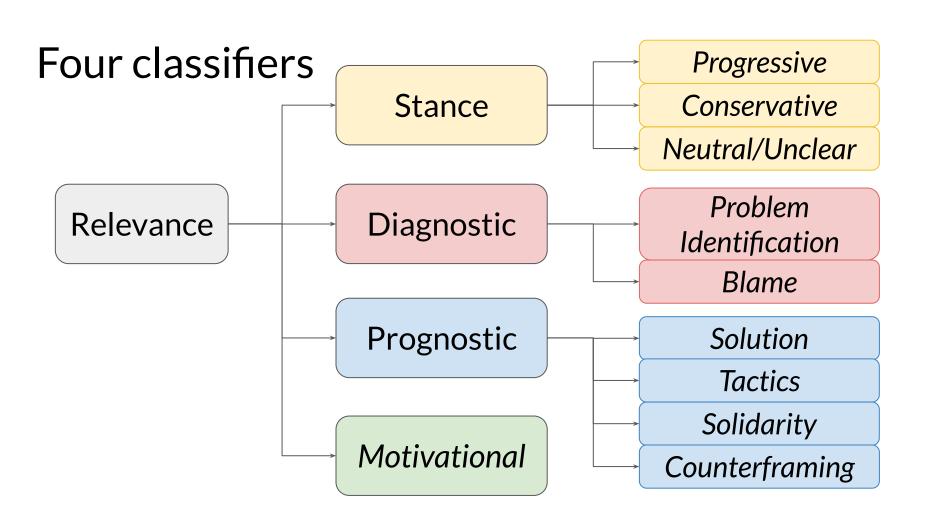
Classification



Analysis: Sociocultural Context



Analysis: Linguistic Properties

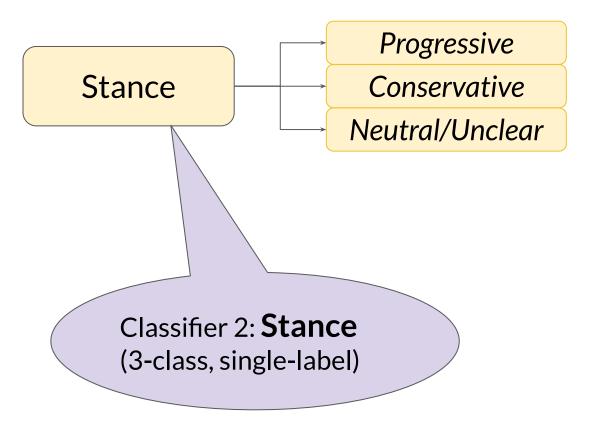


#### Four classifiers

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

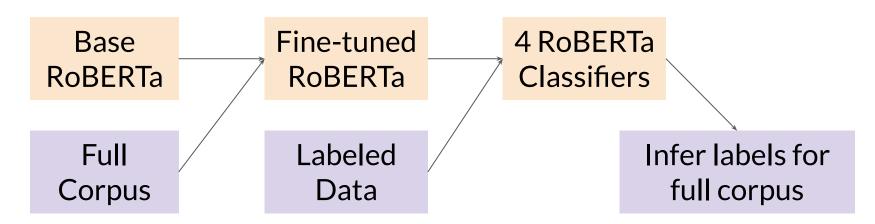
**Tactics** 

Solidarity

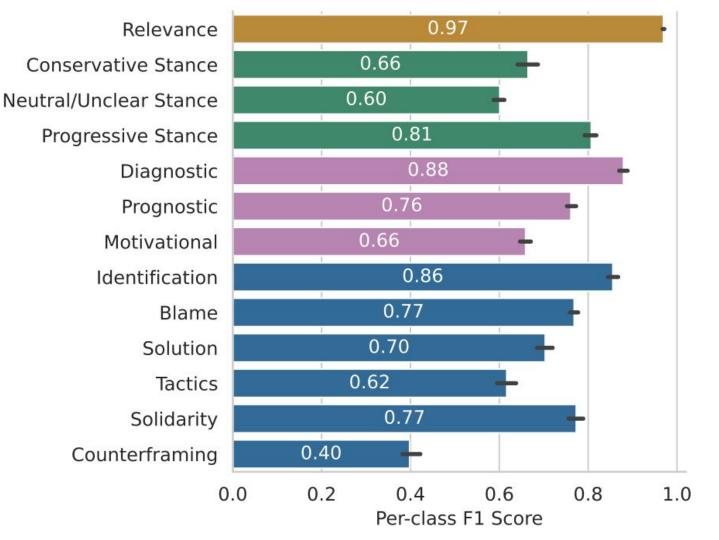
Counterframing

## Model Setup

- Fine-tuned RoBERTa [Liu et al., 2019] on full corpus to better recognize linguistic patterns in social movement tweets
- Trained movement-specific and combined models
- 80% train, 20% test, 5-fold cross-validation



# Model Results





Dataset & Annotation



Classification



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#### Sociocultural Dimensions of Frame Variation

- Movements are situated in time and space
  - Cross-movement diffusion and variation [Carroll & Ratner, 1996]
  - Movement/countermovement frame contests [Stewart et al., 2017]
  - Relationship with offline collective action [Ellingson, 1995]
- Diverse roles and relationships within movements
  - Stakeholder roles (news media, SMOs) [Earl, 2015; Hunt & Gruszczynski, 2023]
  - Target Audiences [Jasper & Poulsen, 1995]
  - Modes of interaction on social media [Garimella et al., 2016]

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Analysis 1: How does attention to each core framing task vary across social and cultural contexts?

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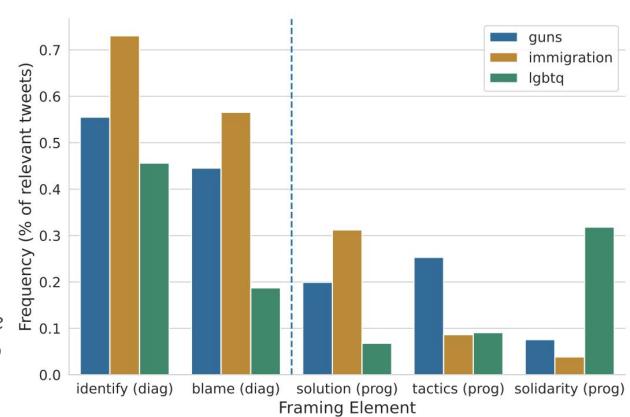
#### Movements attend to different framing elements

 Immigration: diagnostic, solution

Guns: tactics

• LGBTQ: solidarity

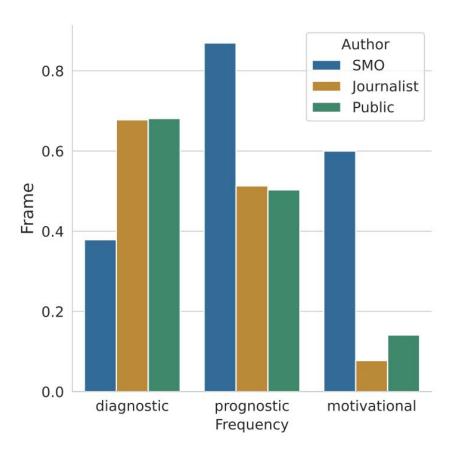
Remains unclear if patterns are stable across time or specific to particular months of data



# Core framing tasks across stakeholder groups

- Labeled authors as journalists, social movement organizations (SMOs), or others (public)
- Out of 1.48M relevant tweets:
- 4,218 relevant tweets from journalists, using existing lists of journalist accounts [Tauberg, 2022; Gotfredsen, 2023]
- 5,817 relevant tweets from SMOs retrieved from *organizer* field of CCC database & Twitter Search API [Bozarth & Budak, 2022]

### More prognostic & motivational framing for SMOs

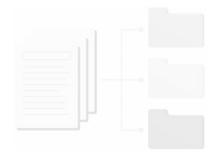


- Are SMOs relevant in the digital age [Earl, 2015]?
- Pattern suggests SMOs play a unique role
- Journalists and public are remarkably similar

# Roadmap



Dataset & Annotation



Classification



Analysis: Sociocultural Context



Analysis: Linguistic Properties

## Identifying linguistic properties in frames

- Log-odds with informative Dirichlet prior for each core framing task within each movement [Monroe et al, 2008]
  - Features statistically overrepresented in one corpus relative to another
- Features: tokens, adjectives, verbs, subject-verb-object
  - SpaCy for tokenization, part-of-speech tagging, and dependency parsing
- Selected top 15 of each feature for qualitative analysis
- Today: a preview of commonalities across movements

Analysis 2: What linguistic features constitute framing tasks?

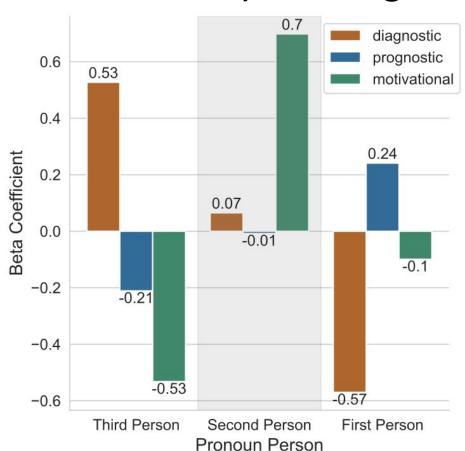
### Pronoun person suggestive of boundary framing

#### From log-odds:

- Diagnostic: 3rd Person (they, their)
- **Prognostic**: 1st Person (*we*, *our*)
- Motivational: 2nd Person (you, your)

#### From logistic regression model:

- Relationship between a pronoun's person and framing tasks in the tweet?
- Fixed Effects: core framing tasks and movements (as controls)
- DVs: pronoun person



### Moral considerations across movements & frames

- Diagnostic: adjectives that expressing moral disapproval
  - Guns: bad, sick, disgusting
  - o Immigration: inhumane, cruel, evil, sick, wrong, disgusting
  - LGBTQ: bad, wrong, disgusting, hateful

- Prognostic: deontic modals of obligation, necessity, intent
  - Need, should, will, must are among the top 15 verbs most associated with tweets with prognostic framing across all three movements.

## Analysis of verbs and their arguments

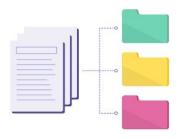
- Violence and neglect are commonly identified problems
  - Violence: Kill, attack, destroy, murder
  - Neglect: fail, ignore, lie, refuse, care, deny, defend

- All movements engage in motivational framing by:
  - Emphasizing the necessity of the movement (need, do\_something)
  - Encouraging readers to join the movement (join, join\_today, join\_us)
  - Pass/support legislation (pass\_legislation, support\_bill, tell\_congress)

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



Dataset & Annotation



Classification



Analysis: Sociocultural Context



Analysis: Linguistic Properties

# Thank you!

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## Extra Slides

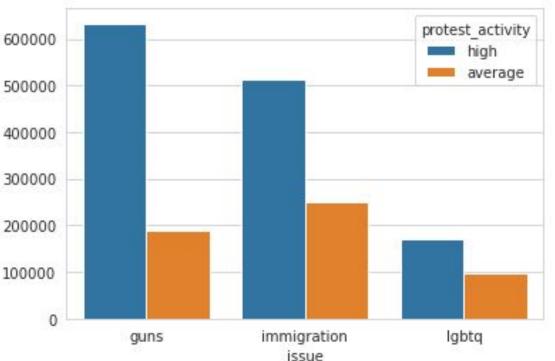
### **Full Dataset Statistics**

1.85M tweets total

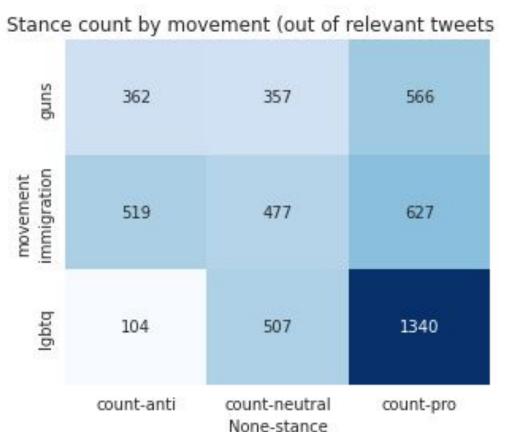
• 822K for guns

• 763K for immigration 400000

• 268K for LGBTQ



### **Stance Distribution**



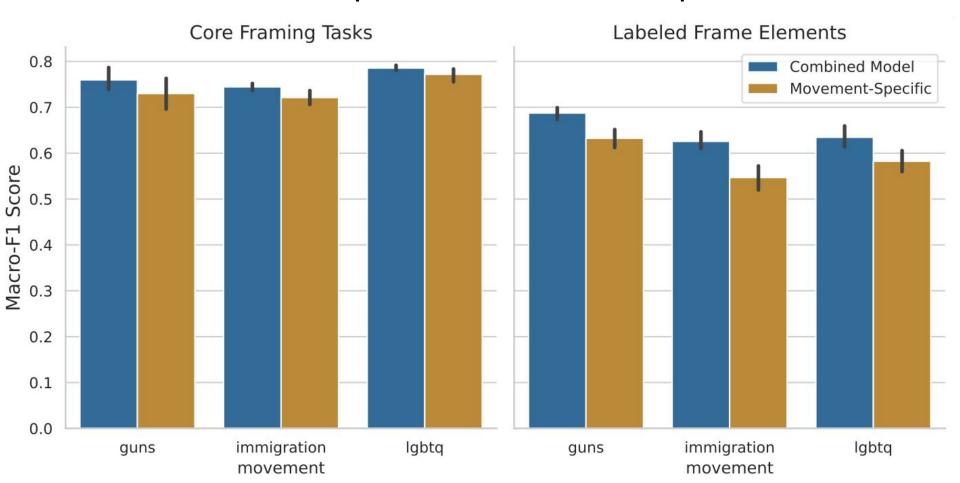
Pro = progressive side (gun control, pro-immigration, pro-LGBTQ)

Anti = conservative side (gun rights, anti-immigration, anti-LGBTQ

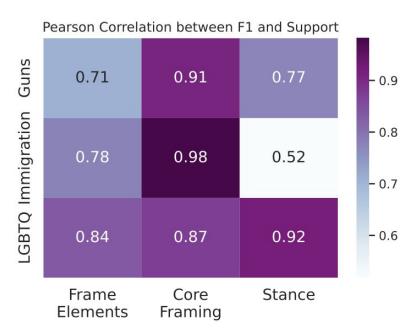
# Frame frequency in annotated dataset

Task	Count	Proportion of total
relevance	4859	0.810
diagnostic	2870	0.478
identify	2655	0.442
prognostic	2143	0.357
blame	1751	0.292
solution	820	0.137
solidarity	737	0.123
motivational	665	0.111
tactics	571	0.095
counter	348	0.058

### Combined model outperforms movement-specific models

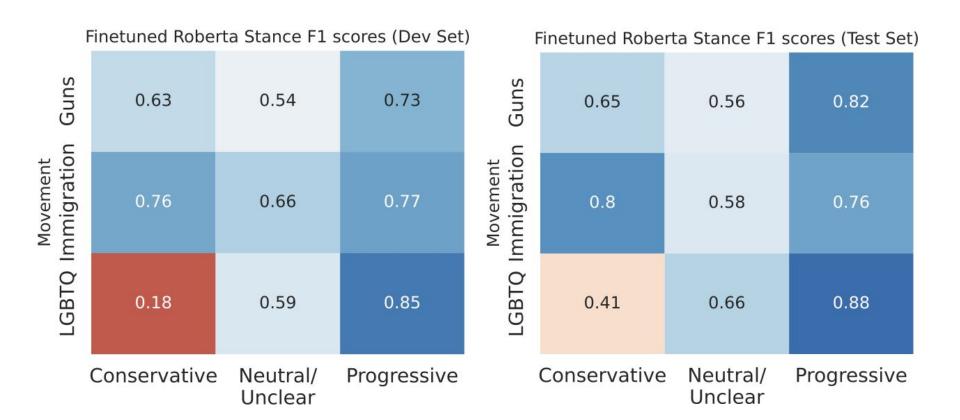


### Data sparsity still hinders model performance



- Pearson correlation between F1 & Support: 0.78
- Potential future work: data augmentation, active learning

### Performance varies across movements (here: stance)



# Movement Frames & Mobilization

Framing Processes

# Core Framing Tasks

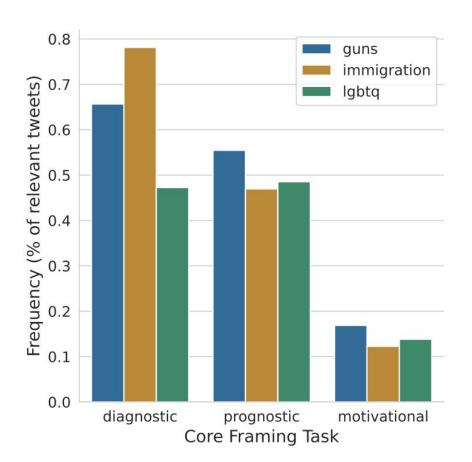
Discursive Processes

**Textual Material** 

Analysis 1: How does attention to each core framing task vary across social and cultural contexts?

Analysis 2: What words, events, topics, and grammatical features constitute each framing task?

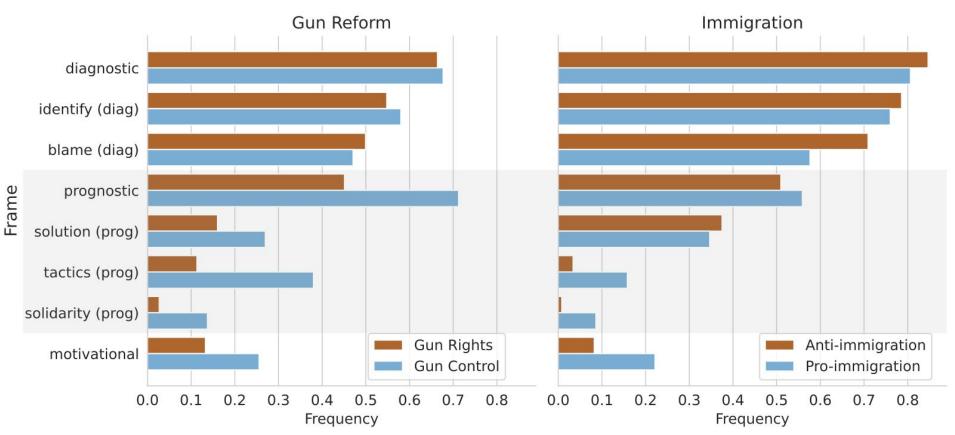
### Cross-movement frame variation



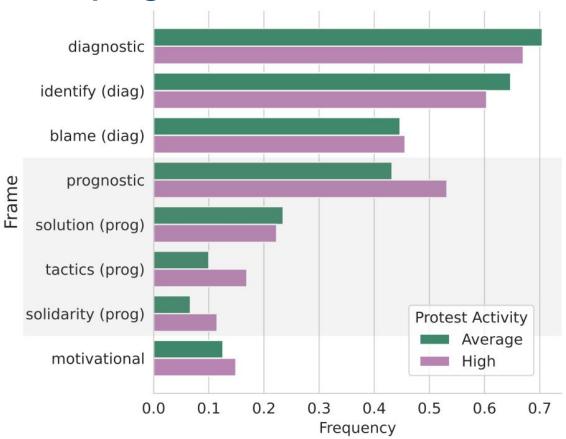
Immigration uses the most diagnostic framing

 Guns uses the most prognostic framing

### Movement/countermovement framing

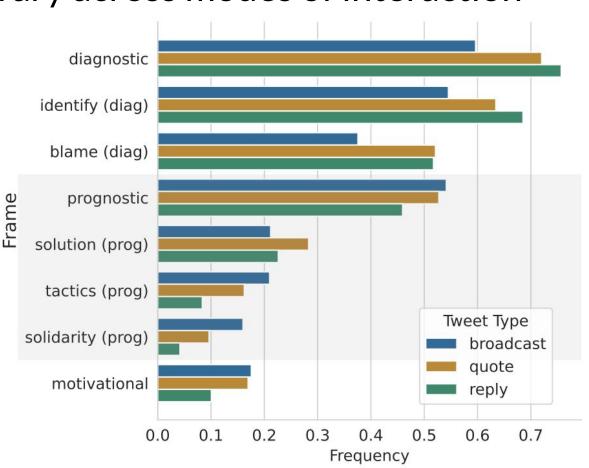


### More prognostic and motivational in high-activity months



### Framing strategies vary across modes of interaction

- Broadcasts, replies, quotes have different intended audiences and discursive goals
- Diagnostic: quotes & replies
- Prognostic & motivational:broadcasts & quotes
- Stable patterns across movements



# More frame alignment for same ideology across movements vs movement/countermovement

- Entropy between distributions of labels (core framing tasks and frame elements) across movement and stance
- Monte Carlo estimation

	movement-stance	Entropy	ci
Shared ideology	Pro-immigration & Pro-gun control	0.0492	0.0002
Shared ideology	Anti-immigration & Anti-gun control	0.0517	0.0001
Shared issue	Pro-gun control & Anti-gun control	0.0898	0.0003
Shared issue	Pro-immigration & Anti-immigration	0.0972	0.0003

	movement group	Cittopy	OI
Lowest frame alignment	(immigration-journalist, immigration-other)	0.024	0.004
between SMOs and	(guns-journalist, guns-other)	0.051	0.005
others	(Igbtq-journalist, Igbtq-other)	0.062	0.007
	(guns-journalist, lgbtq-journalist)	0.085	0.006
	(immigration-journalist, lgbtq-journalist)	0.133	0.008
Highest alignment	(immigration-journalist, guns-journalist)	0.144	0.008
journalists + others	(immigration-smo, lgbtq-smo)	0.153	0.008
within movement	(immigration-smo, immigration-other)	0.193	0.019
	(lgbtq-smo, lgbtq-other)	0.194	0.013
Liliada all'avana antila atore	(guns-smo, lgbtq-smo)	0.196	0.009
High alignment betw	(lgbtq-smo, lgbtq-journalist)	0.213	0.013
journalists across	(immigration-smo, immigration-journalist)	0.247	0.019
movements	(immigration-smo, guns-smo)	0.457	0.021
	(guns-smo, guns-journalist)	0.536	0.020
	(guns-smo, guns-other)	0.593	0.022

movement-group

entropy

ci

### Codebook Diagnostic Examples

Parkland survivor David Hogg to McCain: 'Why do you take so much money from the NRA?'

@realDonaldTrump You are traumatizing children and babies with YOUR policy. End your policy of ripping children and babies from their immigrant parents today and reunify all children and babies with their parents. Let parents out of jail. Seeking asylum not crime.

@KamalaHarris The racist socialist who only cares about illegals and not the American people! #WalkAway #MAGA #LiberalismIsAMentalDisease

It's called denial, these conservative families always deny their own children their right to exist as transgender or gay etc... We (LGBT+) all go through it; abuse, violence & even suicide is all too common, but we are all so engage in ourselves that we hardly feel the pain!

### Codebook Prognostic Examples

The fight for trans equality must be recognised as a class struggle. #Trans people are frozen out of work – this will change only if we move beyond the vicious and obsessive arguments about the boundaries of identity

You know what US students can do besides getting shot and being afraid of getting shot? They MARCH and plan to CHANGE and save people's LIVES. Bravo the Youth #MarchForOurLives

@PhillipLuttazi @RedWaveRising6 @NRA That is why we should stand up for their right to protect themselves, not just let them be victims.

**Happy Pride Month!!!** 

### Codebook Prognostic Examples

@GeorgeTakei Thank you deeply for telling your story on @TheLastWord with Lawrence! You are a hero to many! I'm so sorry for what this nation put you and your family through. #ReunificationNOW #FamiliesBelongTogether �� ② We love you, George!

This is how unspeakable bravery and heroism look. #MarchForOurLives @Emma4Change

@AdvocateProject @SenateDems @realDonaldTrump That is simply not true. There's mountains of evidence to the contrary. Just as the story was women were too emotional to serve - proven false- these stories that transgender persons are too emotional are a load of bullshit.

### Codebook motivational examples

A new poll found that ~69% of Americans think that gun laws should be tightened. ~2% of Americans belong to the #NRA Who do politicians, particularly #Republicans, listen to? Who do they really represent? Let's change that. #Vote2018 #NoGOP #NoNRA

If you're in one of these states trying to plan an event in which CA state employees can participate, ask your state lawmakers to get rid of the laws that enable anti-LGBT discrimination. Even if you DON'T want CA state employees to come to your state, do that.

Please sign this petition for LGBT rights: <URL>

### Common linguistic features of diagnostic framing

- Out-group focus with third-person: they, their, is, are
- Verbs naming violence
  - Kill, attack, destroy, etc. across all movements
  - More speech verbs for LGBTQ (say, call, misgender)
- Verbs suggest that neglecting a problem is a problem itself
  - o Guns: fail, ignore; Immigration: lie, refuse, care; LGBTQ: refuse, deny, defend
- Adjectives that express moral disapproval
  - Guns: bad, sick, disgusting;
  - Immigration: inhumane, cruel, evil, sick, wrong, disgusting
  - o LGBTQ: bad, wrong, disgusting, hateful

### Prognostic framing linguistic features

- Modal verbs of obligation, necessity, and future intent
  - Need, should, will, must are among the verbs most associated with prognostic framing in all three categories
- Subject-verb and verb-object pairs suggest different solutions and tactics across movements
  - Guns: sign\_petition, find\_march, join\_today, ban\_weapon, vote\_them, pass\_legistlation
  - Immigration: build\_wall, stop\_separation, elect\_you, add\_name, secure\_border
  - LGBTQ: celebrate\_pride, show\_support, raise\_awareness

### Motivational framing features

- All movements engage in motivational framing by
  - Emphasizing the necessity of the movement (need, do\_something)
  - Encouraging readers to join the movement (join, join\_today, join\_us)
  - Calls to pass / support legislation (pass\_legislation, support\_bill, tell\_congress)
- But many calls to action vary across movements
  - Voting for guns and immigration, not LGBTQ
  - Education for LGBTQ and immigration, not guns
  - Donation for LGBTQ and immigration, not guns