# A Holistic Analysis Framework for Analyzing the Covid-19 Vaccine Debate

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#### **Our Work:**

VACCINE

Characterizing the *reasons* and attitudes that guide people's choices

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Stance: Not a fan

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**Why?** The vaccine is dangerous

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**Why?** The vaccine is dangerous

Stance: Not a fan

The vaccine framed as *harmful* to *pregnant women* 

#### **An Interconnected View of Opinions**

I never saw anything like this **Government's** obsession with <u>citizens</u> getting the COVID vaccine. I look at the actions **Biden** is willing to do to <u>us</u> and it makes me refuse to get the shot even greater. Is this a trial run a **Socialist dictatorship**???

#### **Opinion** Analysis

Vaccination Stance: Negative Reason: Government distrust

#### **Morality Frame Analysis:**

Moral Foundation: Oppression Negative Actors: Government, Biden, Socialist dictatorship Negative Targets: citizens, us

Identifying Morality Frames in Political Tweets Using Relational Learning, [Roy, Pacheco and Goldwasser, 2021]

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• Pro or anti vaccine, and the *reasons why* 



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Big challenge

Interactive, humans-in-theloop protocol

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- First session: adding new themes, removing themes that were not prevalent in the data
- Second session: identifying high-level argumentative patterns that were not being captured, and contributing 2-5 examples

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If the actor of AUTHORITY is "Fauci", the author is more likely to be pro-vaccine, and to express Trust in Science

If the actor of HARM is "Fauci", the author is more likely to be anti-vaccine, and to express Distrust in Government



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- Entities relations and prob. rules
- Learn embeddings for ent/rels
- Learn weights for rules
- Constrained optimization

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 We beat the fully supervised base model with 50% of direct supervision

# The Impact of Reasons and Interaction



### + all other rules

### **Is Human Intervention Helpful?**

### • Stronger correlations!

#### Correlation heatmaps between **Themes** and **Moral Foundations**



LDA Topics

#### Wawrutza et. al, 2021 Themes

#### **Interactive Themes**

## **Is Human Intervention Helpful?**

• More diverse distribution of reasons!



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- Structured inference and human interaction are good alternatives