Overview Data and Methodology Results Conclusion

# Analyzing Voter Sentiment through 2020 U.S. Election Twitter Data Text Analysis Project

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## Overview

- ► Motivation: How to understand political polarization and voter sentiment with social media comments?
- ► Research Question: how voter sentiment and Twitter language differ for different political stances and political topics.
  - 1. Predicting political stance
  - 2. Topic modelling
  - 3. Sentiment analyis
- ► Finding: Most Twitter sentiments are polarized and tend to be negative.
  - Censorship scandal and Social Security: almost no positive sentiments
  - Biden-learning tweets talk more about the Final Presidential Debate and Racism; Trump-leaning tweets talk more about censorship and ballot polls.

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# Data and Methods

- Data Sources:
  - Kaggle's U.S. ELection 2020 Dataset, cleaned to a remaining 266k tweets from period 10/15-11/18.
  - Georgetown University's 'Knowledge Enhance Masked Language Model for Stance Detection'. 2500 tweets pre-labeled with stance
- Methods and Evaluation:
  - Multinomial Logit-lasso regression: group to 3 dummies, for-Trump/against-Biden, neutral, for-Biden/against-Trump
  - LDA: setting to 20 political topics, spit out key words identified to topics like censorship and social security
  - Huggingface's Tweet-roBERTa algorithm: identify the percentages of positive, neutral, and negative language use.

# Results

### Multinomial Logit accuracy score of 0.779

|    | word                  | coef      |
|----|-----------------------|-----------|
| 1  | maga                  | -6.584186 |
| 2  | democrat              | -4.653874 |
| 3  | kag                   | -3.281676 |
| 4  | red                   | -1.999995 |
| 5  | best                  | -1.795547 |
| 6  | americafirst          | -1.578382 |
| 7  | makeamericagreatagain | -1.113084 |
| 8  | year                  | -1.071948 |
| 9  | new                   | -0.876847 |
| 10 | try                   | -0.805163 |

Table 1: Top 10 words predictive of Trump-supporting/Biden-opposing

| 1         votebluetosaveamerica         17.437117           2         voteblue         10.586887           3         dumptrump         6.271288           4         demonvention         5.228917           5         gop         3.796428           6         votebluenomatterwho         3.787612           7         joebiden         2.492431           8         trump         2.22982           9         votebluetoendthisnightmare         2.15652 |    |   |           |
|--|----|---|-----------|
| 2         voteblue         10.586887           3         dumptrump         6.271288           4         demconvention         5.228917           5         gop         3.796428           6         votebluenomatterwho         3.787613           7         joebiden         2.4924318           8         trump         2.229823           9         votebluetoendthisnightmare         2.156523   |    | word                                    | coef      |
| 3         dumptrump         6.271288           4         demconvention         5.22891           5         gop         3.75642           6         voteblenomatterwho         3.78761           7         joebiden         2.492431           8         trump         2.22982           9         votebluetoendthisnightmare         2.15652   | 1  | votebluetosaveamerica                   | 17.437117 |
| 4 demonvention 5.228915 5 gop 3.706429 6 votebhenomatterwho 3.787617 7 joebiden 2.492431 8 trump 2.229826 9 votebhetoendthisnightmare 2.156527   | 2  | voteblue                                | 10.586887 |
| 5         gop         3.796428           6         votebluenomatterwho         3.787613           7         joebiden         2.492431           8         trump         2.228823           9         votebluetoendthisnightmare         2.156523   | 3  | dumptrump                               | 6.271283  |
| 6         votebluenomatterwho         3.787613           7         joebiden         2.492431           8         trump         2.229823           9         votebluetoendthisnightmare         2.156523  | 4  | demconvention                           | 5.228915  |
| 7 joebiden 2.492431<br>8 trump 2.229823<br>9 votebluetoendthisnightmare 2.156523   | 5  | gop                                     | 3.796428  |
| 8 trump 2.229823<br>9 votebluetoendthisnightmare 2.156522  | 6  | votebluenomatterwho                     | 3.787613  |
| 9 votebluetoendthisnightmare 2.156522  | 7  | joebiden                                | 2.492431  |
|  | 8  | trump                                   | 2.229823  |
| 10 debate 2.074056   | 9  | ${\it voteblue to end this night mare}$ | 2.156522  |
| 2011000  | 10 | debate                                  | 2.074056  |

Table 2: Top 10 words predictive of Biden-supporting/Trump-opposing

#### ▶ 5 main topics

Table 4: Top words of the most mentioned political topics.

|                 | Racism(14) | Media Censorship(13) | Presidential Debate(7) | Election/Ballot Progress(2) |
|-----------------|------------|----------------------|------------------------|-----------------------------|
| Tax             | President  | Medium               | Watch                  | election2020                |
| Healthcare      | Trump      | hunterbiden          | Townhall               | Arizona                     |
| Democracy       | Black      | Twitter              | ABC (news)             | GA (Georgia)                |
| Pay             | Woman      | censorship           | Tonight                | Pennsylvania                |
| Social Security | Racism     | News                 | trumpmeltdown          | everyvotecounts             |
| Election        | Vote       | Ukraine              | President              | campaign                    |

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# Final Results and Conclusion

- ► Biden supporters tweet more about the final Presidential debate
- Trump supporters tweet more about Biden's media censorship scandal and ballot progress
- Very low share of positive tweets in clearly Trump and Biden camp tweets
- Biggest share of negative tweets in media censorship

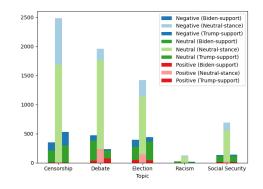


Figure: Three bars from left to right: Biden-support, neutral, Trump support. Stacked bars from top to bottom: negative, neutral, positive