

Shifting Trends of COVID-19 Tweet Sentiment with Respect to Voting Preferences in the 2020 Election Year of the United States

Megan Doman, Jacob Motley, Hong Qin, Mengjun Xie, Li Yang
Department of Computer Science and Engineering
The University of Tennessee at Chattanooga
Chattanooga, Tennessee, United States

Presenter: Hong Qin
hong-qin@utc.edu

HUSO 2022, Contribution 80021



Hong Qin

Associate Professor, Department of Computer Science and Engineering, Department of Biology, Geology and Environmental Science at the University of Tennessee at Chattanooga. He previously worked at Spelman College, Tuskegee University, and University of Rochester. He completed his postdoc training and Ph.D. at the University of Chicago.



Introduction



U.S. Politics & Covid-19

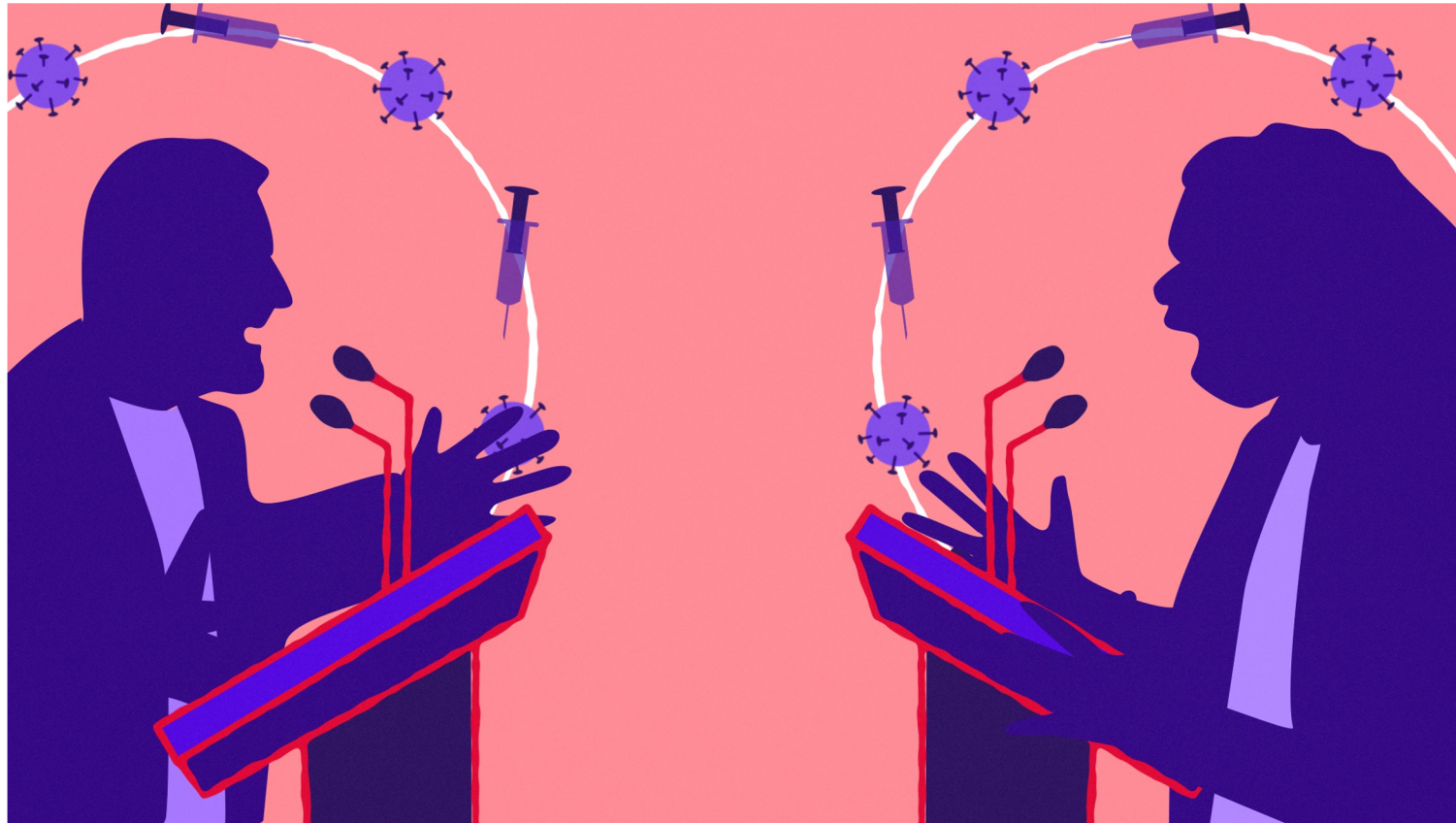
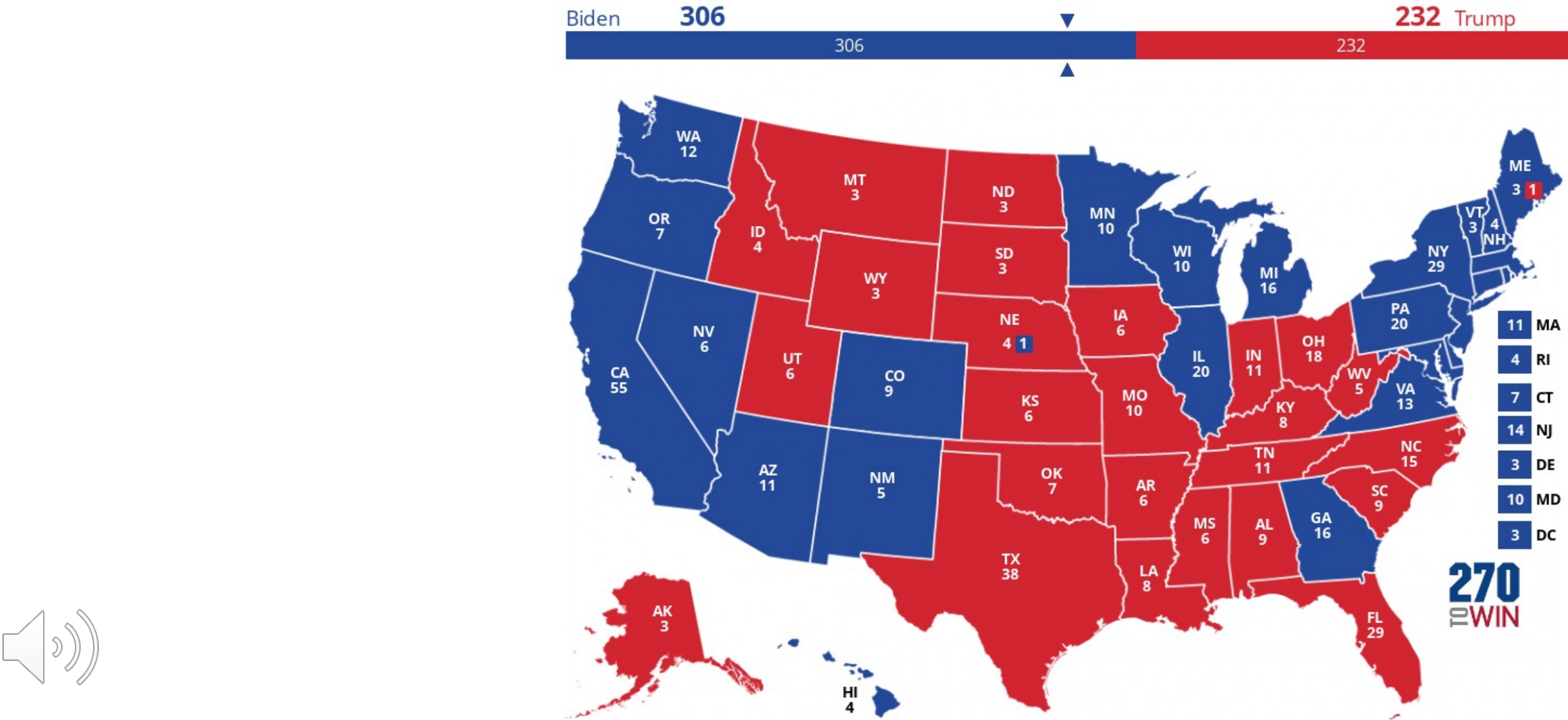


Image source: [statnews.com](https://www.statnews.com)



Blue and Red States in the U.S.A.



Twitter as a Data Source

- API for app development
- Tweet objects and metadata:
 - Date
 - Location
 - Tweet contents
- Dataset of Tweets based on 90 key words surrounding Covid-19



Dr. Marvin Lerner M.D., Ph.D.
@DrMLernerMDPhD



New covid drug Paxlovid pushed and financed by Government does not work. Medical Malpractice. Ivermectin now available in New Hampshire and Tennessee is proven effect Not approved by big pharma because it is available and cheap



wsj.com

Pfizer's Covid-19 Pill Failed Study Testing Its Preventative Use

Paxlovid, authorized to treat people at high risk of severe disease within a few days of infection, wasn't found to prevent symptomatic infections in a study ...

9:21 PM · Apr 29, 2022 from New York, USA · Twitter for Android

14 Retweets 4 Quote Tweets 29 Likes



Sentiment Analysis & NLP

Sentiment scores ~ Mood

😞 $[-1, 1]$ 😊



Goals

Is there association
between COVID-19
tweets and voting
preferences in the
2020 U.S.A. election?



Tweet Data

➤ [Appl Intell \(Dordr\)](#). 2021;51(5):2790-2804. doi: 10.1007/s10489-020-02029-z. Epub 2020 Nov 6.

Design and analysis of a large-scale COVID-19 tweets dataset

Rabindra Lamsal ¹

Table 1 Overview of the filtering keywords as of July 17, 2020

In use since	Keywords ^a
March 20, 2020	corona, #corona, coronavirus, #coronavirus
April 18, 2020	covid, #covid, covid19, #covid19, covid-19, #covid-19, sarscov2, #sarscov2, sars cov2, sars cov 2, covid_19, #covid_19, #ncov, ncov, #ncov2019, ncov2019, 2019-ncov, #2019-ncov, #2019ncov, 2019ncov
May 16, 2020	pandemic, #pandemic, quarantine, #quarantine, flatten the curve, flattening the curve, #flatteningthecurve, #flattenthecurve, hand sanitizer, #handsanitizer, #lockdown, lockdown, social distancing, #socialdistancing, work from home, #workfromhome, working from home, #workingfromhome, ppe, n95, #ppe, #n95



Sample Tweets	Date, Location	Sentiment
Dinner tonight is a B.L.T. and a corona. Sponsored by Covid-19 @ Miami Beach, Florida https://t.co/GyznR8Yqed	2020-04-11 Miami Beach, FL	0
Bare shelves: the time of #corona @ South Burlington, Vermont https://t.co/idfYhEOffV	2020-04-11 South Burlington, VT	0
EMBRACE THE SUCK: In spirituality, contrasting experiences refers to the ones we'd really rather not have. Maybe it's seasonal chronic illness woes (👩\u200d♀), perhaps it's various elements of Corona chaos, or it could https://t.co/upvcbRBAZS	2020-04-10 Long Beach, CA	-0.87
The best alone time I can have. Wanted to share 🙏\n#stayhome #staysafe #socialdistancing #stayhomestaysafe #inspiring #fun #creative #art #collab #teamwork #newmusic #newalbum #cancel #corona #solo #guitar #life https://t.co/edpaPmXrHa	2020-04-10 Los Angeles, CA	0.91
This quarantine turning me into Joe Jackson 🧑♂️😂😂😂\n.\n.\n.\n#llfruge #tallformyheight #mvpofnochildleftbehind #quarantinedad #jacksons #beyonce #Brooklyn #corona #corona #covid #fatherdaughter #florida @ Fleming https://t.co/xmAHSg39hZ	2020-04-11 Fleming Island, FL	0.83



Tweet Sentiment Estimation

- VADER Sentiment
 - Designed for social media
- Average by state per day

date	AK	AL	AR	AZ	CA	CO	CT	DC	DE	FL	GA	HI	IA	ID	IL	IN	KS
20200320	0	0.8591	0.128	0.4961	0.159037	0.632033	0.13844	0	0	0.236633	0.128652	0	0	0	0.22034	0.269864	0
20200321	0	0.550667	0.2202	0.34802	0.32214	0.4679	0	0	0	0.287658	0.275525	0	0.8553	0	0.194075	0.291994	
20200322	0	0.471225	0	0.612588	0.253292	0.291575	0.7845	0	0.5409	0.522935	0.463239	0	0	0	0.423533	0.149189	0
20200323	0	0	0.631	0.560175	0.274739	0.386767	0.7269	0	0	0.143119	0.129638	0	0.8807	0	0.1642	0.352373	
20200324	0	0	-0.05615	0.27142	0.286396	0.34	0.9287	0	0	0.281813	0.251791	0	0.03545	0	-0.04134	0.219129	
20200325	0	0	0.6632	0	0.151723	0.6369	0.1635	0	0	-0.05179	0.197659	0	0	0.4295	0.085583	0.15485	0
20200326	0	-0.4341	0	-0.2846	0.286078	0.308	0.33985	0	-0.2263	0.2266	0.25732	0	-0.1079	0	0.65615	0.190742	0.3
20200327	0	0.2263	0	0.41085	0.126934	0.23575	0.440967	0	0	0.287253	0.166187	0	0.29295	0	0.250567	0.473975	0
20200328	0	0	-0.0772	0.55085	0.304568	0.362233	0	0	0	0.399	0.109365	0	0.6379	0	0.2844	0.104029	0.44
20200330	0	0	0	0.1289	0.211116	0.371967	0	0	0	0.347369	0.246344	0	0.4215	-0.5423	-0.01425	0.27855	
20200331	0	0.2202	0.2202	0.557225	0.182626	0.095633	0	0	0.128	0.266181	0.2041	0	0.4753	0.10115	0.520425	0.200633	
20200401	0	-0.1697	0	0.21665	0.327692	-0.3612	0.3243	0	0	0.359375	0.457409	0	0.6705	0.8126	0.4559	-0.31705	0.4
20200402	0	0	0.4215	0.414683	0.266767	0.29352	0.855	0	0	0.422056	0.423283	0	-0.7351	0	0.4441	0.206836	
20200403	0	0	0.80705	0.446233	0.204173	0.314367	0.3556	0	0	0.342	0.2579	0	0.6808	0	0.4404	0.5098	
20200404	0	0.1543	0	0.237125	0.345734	0.15835	0	0	0	0.125864	0.324313	0	0.6239	0	0.8516	0.556686	
20200405	0	0	0.8832	0.21116	0.278304	0.046867	0.4019	0	0	0.260831	0.470071	0	-0.0292	0	0.8385	0.146643	
20200406	0	0.492967	-0.08985	0.4286	0.309959	-0.02548	0.6369	0	0	0.152573	0.606243	0	0	0.3612	0.33765	0.626186	
20200407	0	0.7783	0	-0.2553	0.249727	0.362967	0	0	0	0.436467	0.277655	0	0	0.7545	0.36206	0.468656	
20200408	0	0.9473	0	-0.5252	0.11149	0.4654	0	0	0	0.459817	0.281789	0	0	0	0.09636	-0.05503	
20200409	0	0	0	0.401367	0.249562	-0.4291	0	0	0	0.10918	0.1913	0	0	0	0.9274	0.31948	
20200410	0	0	0	0.2295	0.280111	0.190633	0	0	0	0.204125	0.2951	0	0	0	-0.5423	-0.02277	



Voting Data

- State, party called, percent

state	called	final	dem_votes	rep_votes	other_votes	dem_percent	rep_percent	other_percent	dem_this	margin_sh	vote_chan	stateid
U.S. Total	D	Yes	81,282,916	74,223,369	2,891,441	51.30%	46.90%	1.80%	4.50%	2.40%	15.90%	
15 Key Battlegrounds			31,908,248	33,002,287	950,354	48.40%	50.10%	1.40%	-1.70%	2.00%	17.20%	
Non-Battlegrounds			49,374,668	41,221,082	1,941,087	53.40%	44.50%	2.10%	8.80%	2.70%	15.10%	
15 Key Battlegrounds												
Arizona	D	Yes	1,672,143	1,661,686	53,497	49.40%	49.10%	1.60%	0.30%	3.90%	31.60%	AZ
Florida	R	Yes	5,297,045	5,668,731	101,680	47.90%	51.20%	0.90%	-3.40%	-2.20%	17.50%	FL
Georgia	D	Yes	2,473,633	2,461,854	62,229	49.50%	49.30%	1.20%	0.20%	5.40%	22.10%	GA
Iowa	R	Yes	759,061	897,672	34,138	44.90%	53.10%	2.00%	-8.20%	1.20%	8.00%	IA
Maine 2nd District	R	Yes	168,696	196,692	10,961	44.80%	52.30%	2.90%	-7.40%	2.80%	6.50%	ME2
Michigan	D	Yes	2,804,040	2,649,852	85,410	50.60%	47.80%	1.50%	2.80%	3.00%	15.40%	MI
Minnesota	D	Yes	1,717,077	1,484,065	76,029	52.40%	45.30%	2.30%	7.10%	5.60%	11.30%	MN
Nebraska 2nd District	D	Yes	176,468	154,377	6,909	52.20%	45.70%	2.00%	6.50%	8.80%	15.80%	NE2
Nevada	D	Yes	703,486	669,890	32,000	50.10%	47.70%	2.30%	2.40%	0.00%	24.90%	NV
New Hampshire	D	Yes	424,921	365,654	15,607	52.70%	45.40%	1.90%	7.40%	7.00%	8.30%	NH
North Carolina	R	Yes	2,684,292	2,758,775	81,737	48.60%	49.90%	1.50%	-1.30%	2.30%	16.50%	NC
Ohio	R	Yes	2,679,165	3,154,834	88,203	45.20%	53.30%	1.50%	-8.00%	0.10%	7.70%	OH
Pennsylvania	D	Yes	3,458,229	3,377,674	79,380	50.00%	48.80%	1.10%	1.20%	1.90%	13.10%	PA
Texas	R	Yes	5,259,126	5,890,347	165,583	46.50%	52.10%	1.50%	-5.60%	3.40%	26.20%	TX
Wisconsin	D	Yes	1,630,866	1,610,184	56,991	49.40%	48.80%	1.70%	0.60%	1.40%	10.80%	WI
Non-Battlegrounds												
Alabama	R	Yes	849,624	1,441,170	32,488	36.60%	62.00%	1.40%	-25.50%	2.30%	9.40%	AL
Alaska	R	Yes	153,778	189,951	15,801	42.80%	52.80%	4.40%	-10.10%	4.70%	12.80%	AK
Arkansas	R	Yes	423,932	760,647	34,490	34.80%	62.40%	2.80%	-27.60%	-0.70%	7.80%	AR
California	D	Yes	11,110,250	6,006,429	384,192	63.50%	34.30%	2.20%	29.20%	-0.90%	23.40%	CA
Colorado	D	Yes	1,804,352	1,364,607	87,993	55.40%	41.90%	2.70%	13.50%	8.60%	17.10%	CO
Connecticut	D	Yes	1,080,831	714,717	28,309	59.30%	39.20%	1.60%	20.10%	6.40%	10.90%	CT

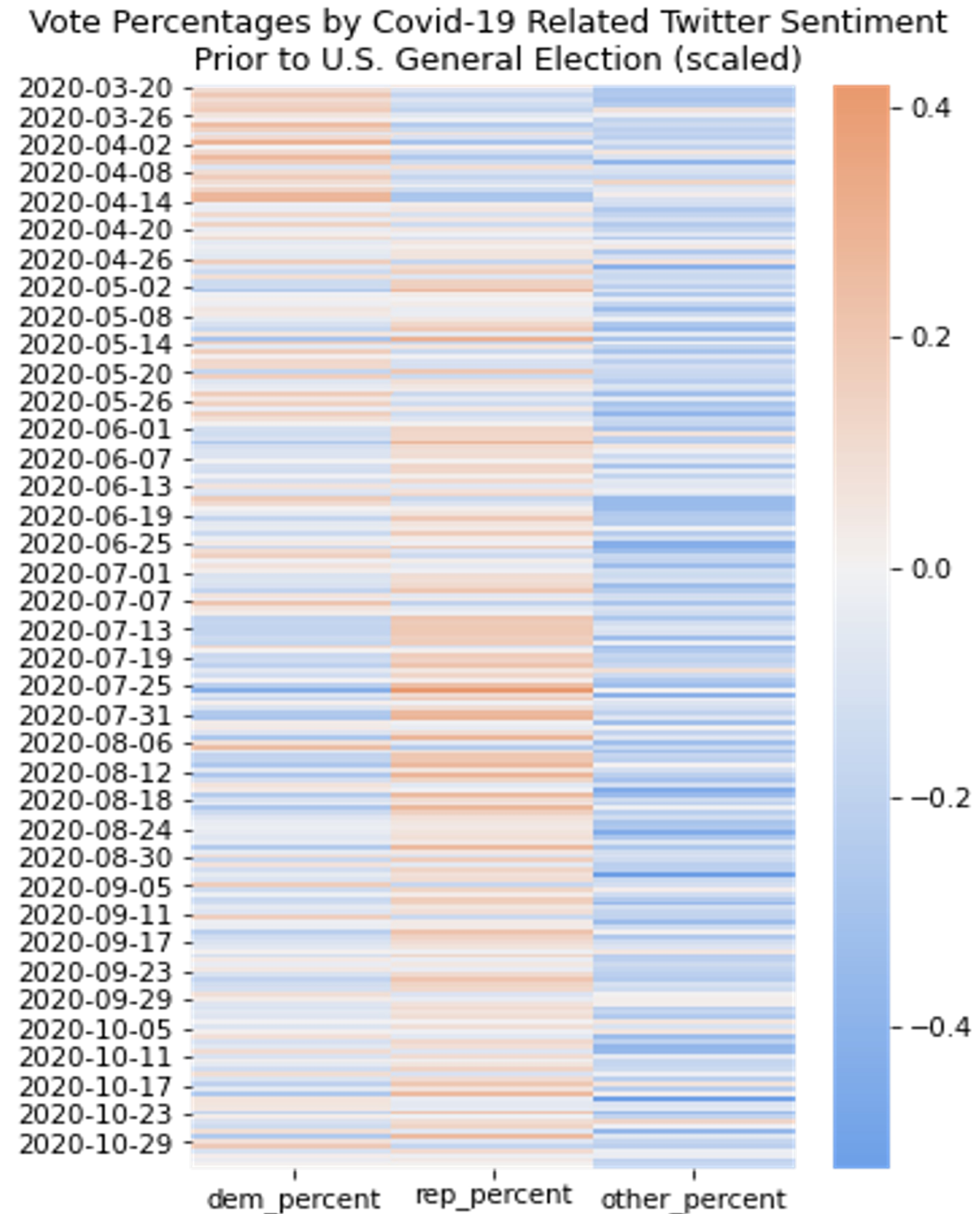


Results



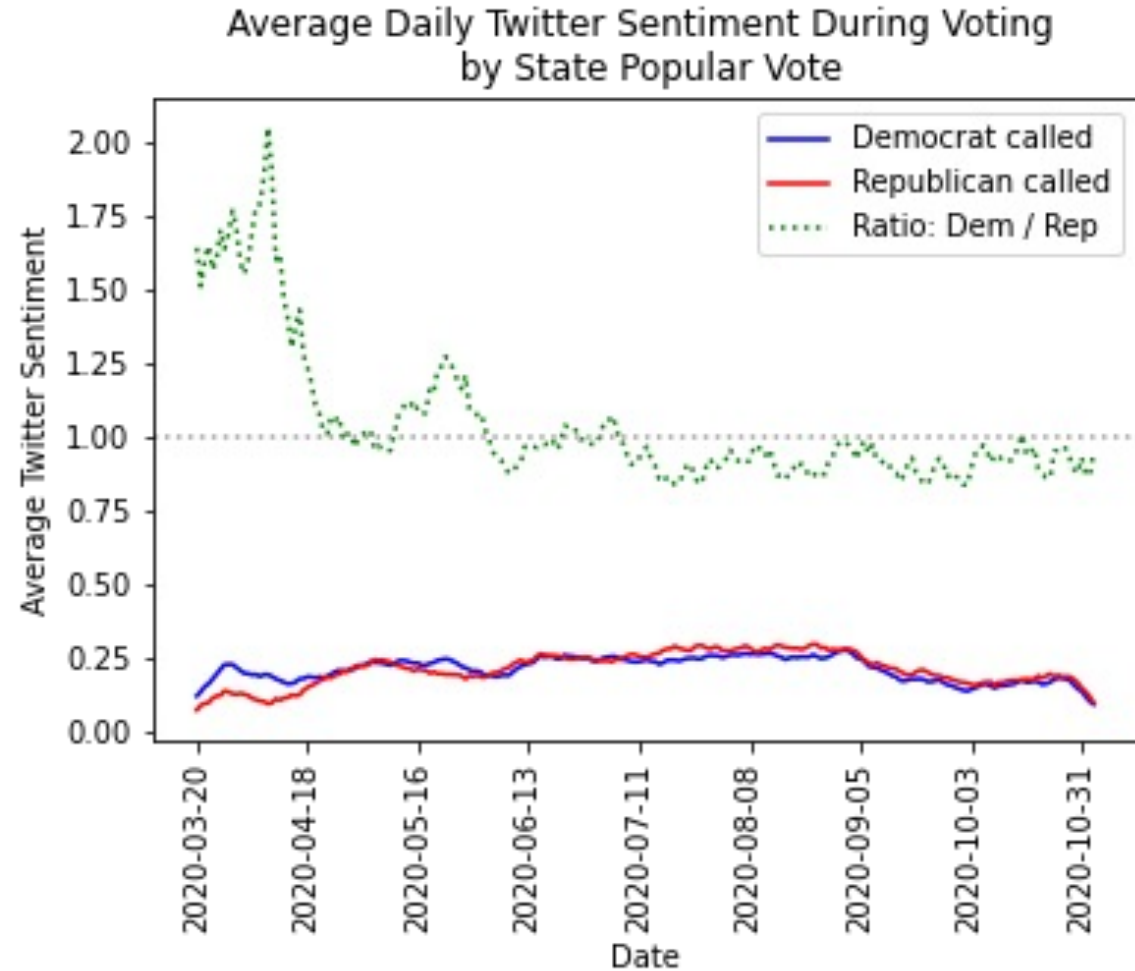
Correlation Heat Map

Daily average tweet sentiment
~ votes count by state



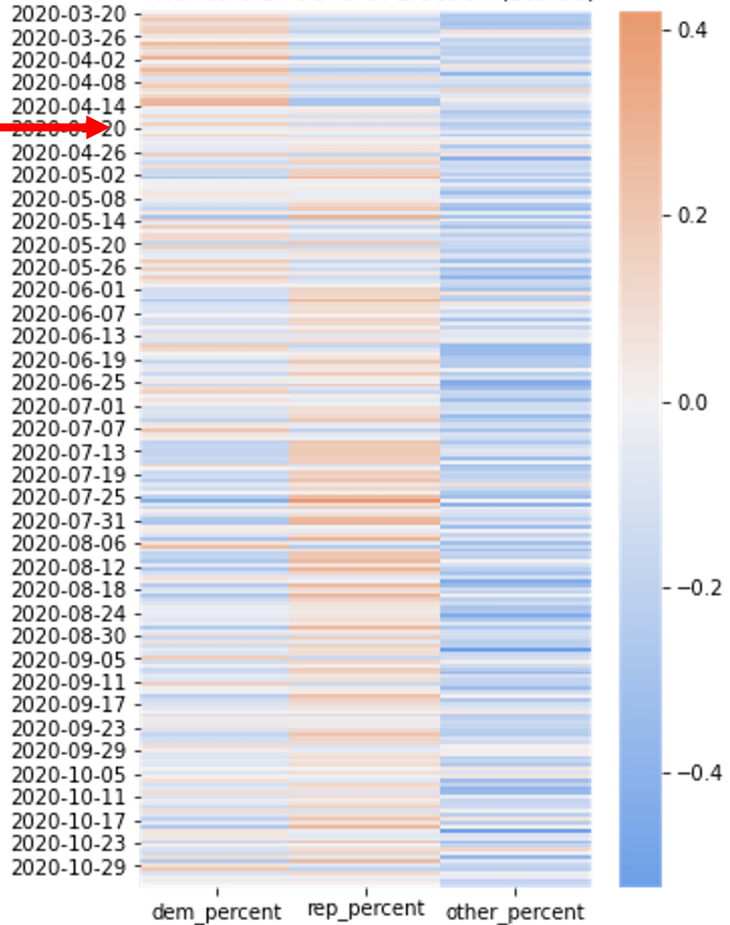
Time Series

- Sentiment Averages for states by party called
- Ratio:
 - Avg. Dem Sentiment $\div$ Avg. Rep Sentiment

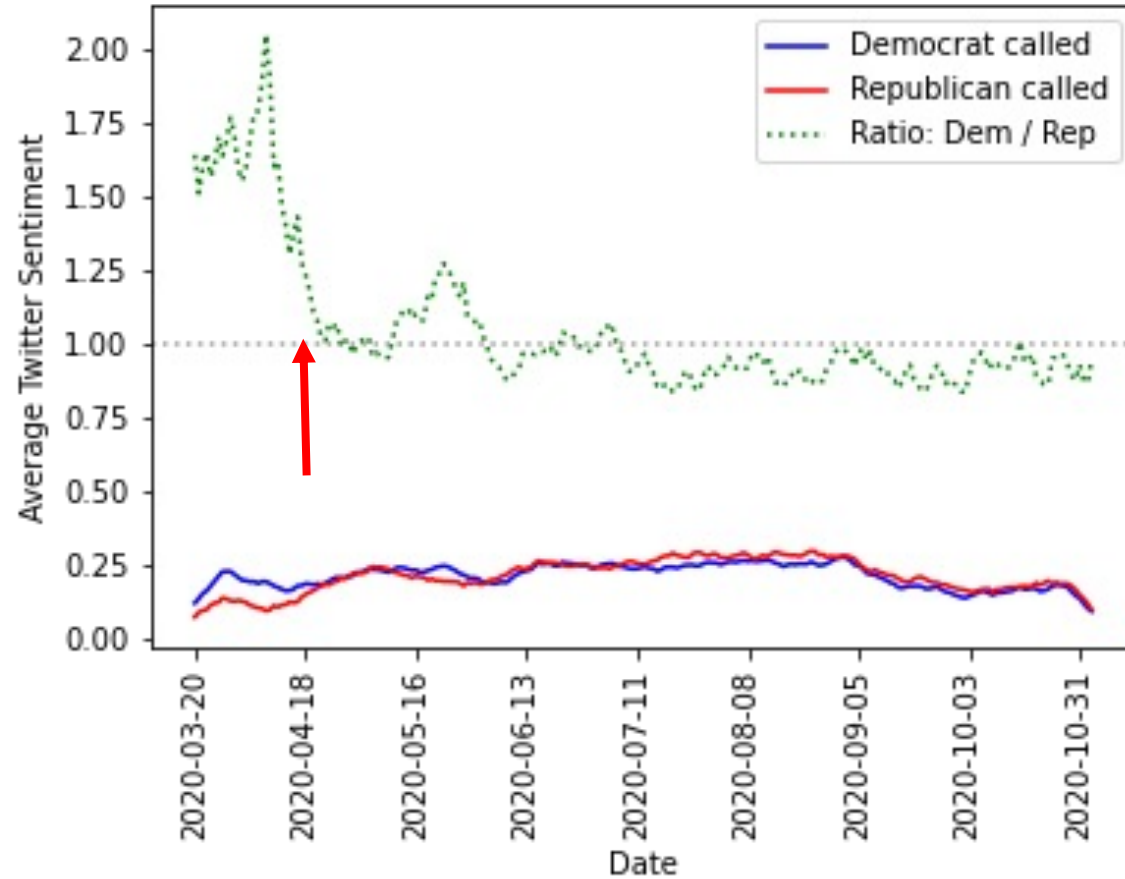


Trends

Vote Percentages by Covid-19 Related Twitter Sentiment
Prior to U.S. General Election (scaled)



Average Daily Twitter Sentiment During Voting
by State Popular Vote



Conclusion

We observed weak correlations between the average sentiment of COVID-19 related tweets and popular votes in two-week intervals, and the trends gradually become opposite for Democratic and Republican voting preferences. We then compared the average sentiments of COVID-19 related tweets between states called in favor of the Republican (red states) or Democratic parties (blue states). We found that at the beginning of lockdowns COVID-19 tweet sentiments in the blue states were much more positive than those in the red states. However, COVID-19 tweet sentiments in the red states gradually become more positive during the summer of 2020 and persisted until the election day.



Acknowledgements

We are thankful for the support of NSF BD Spoke #1761839, SFS #1663105, REU #1852042 and #2149956, NSF REU #1852042 and #2149956

An internal CEACSE award and support from the Office of the Vice Chancellor of Research at the University of Tennessee at Chattanooga.



Sources

- [1] Gusmano, M.K., et al., *Partisanship in initial state responses to the COVID - 19 pandemic*. World Medical & Health Policy, 2020. 12(4): p. 380-389.
- [2] Harvey, N., *Behavioral fatigue: Real phenomenon, naïve construct, or policy contrivance?* Frontiers in Psychology, 2020: p. 2960.
- [3] Liu, H., et al., *COVID-19 information overload and generation Z's social media discontinuance intention during the pandemic lockdown*. Technological Forecasting and Social Change, 2021. 166: p. 120600.
- [4] Burel, G., T. Farrell, and H. Alani, *Demographics and topics impact on the co-spread of COVID-19 misinformation and fact-checks on Twitter*. Information Processing & Management, 2021. 58(6): p. 102732.
- [5] Battaglio, S., *When red meant Democratic and blue was Republican. A brief history of TV electoral maps.*, in *Los Angeles Times*. 2016.
- [6] Wang, H., et al. *A system for real-time twitter sentiment analysis of 2012 us presidential election cycle*. in *Proceedings of the ACL 2012 system demonstrations*. 2012.
- [7] Lamsal, R., *Design and analysis of a large-scale COVID-19 tweets dataset*. Applied Intelligence, 2021. 51(5): p. 2790-2804.
- [8] Staff, A., *A Timeline of COVID-19 Developments in 2020*. American Journal of Managed Care, 2021. 1.
- [9] Wasserman, D., et al., *2020 Popular Vote Tracker*. 2021: Cook Political.
- [10] Hutto, C. and E. Gilbert. *Vader: A parsimonious rule-based model for sentiment analysis of social media text*. in *Proceedings of the international AAAI conference on web and social media*. 2014.
- [11] Waskom, M.L., *seaborn: statistical data visualization*. Journal of Open Source Software, 2021. 6(60): p. 3021.
- [12] Mislove, A., et al., *Understanding the Demographics of Twitter Users*. Proceedings of the International AAAI Conference on Web and Social Media, 2021. 5(1): p. 554-557.