

We have to stop
investing in
fossil fuels!

#divest

Warmest
year in
decades!

DSL

Chemical Spills and Environmental Anxiety on Social Media: Linking Twitter Data with National Survey, 2010-2023

No deep
killing

CO2 cuts
by 2020!

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April. 28. 2023

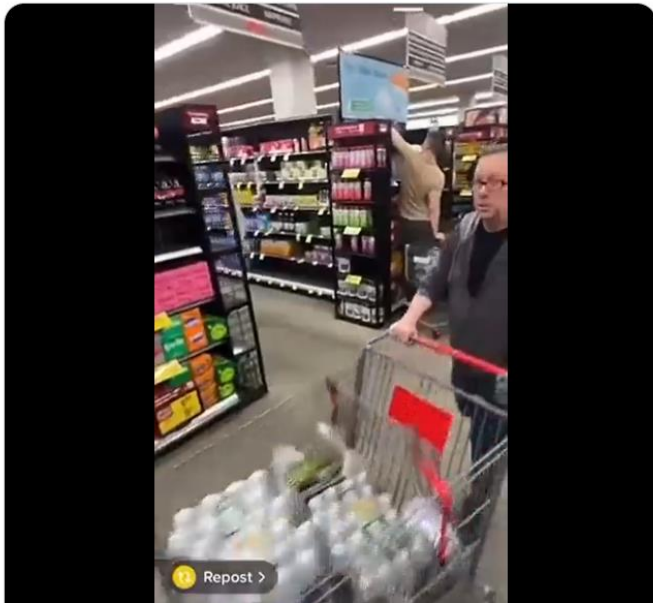


Mrgunsgear @Mrgunsgear · Mar 26

Unprepared **Philly** residents scrambling for **water** after the chemical spill in Bristol, PA....

Plan accordingly....

#Prepare #water #ready #prepper #bristol #philly #Pennsylvania #ChemicalSpill #Latex



2



Charlie Flowe @DjFlowe · Mar 26

Me... after drinking **Philly tap water** at 12:01 AM Tuesday.
#philadelphia



8

161

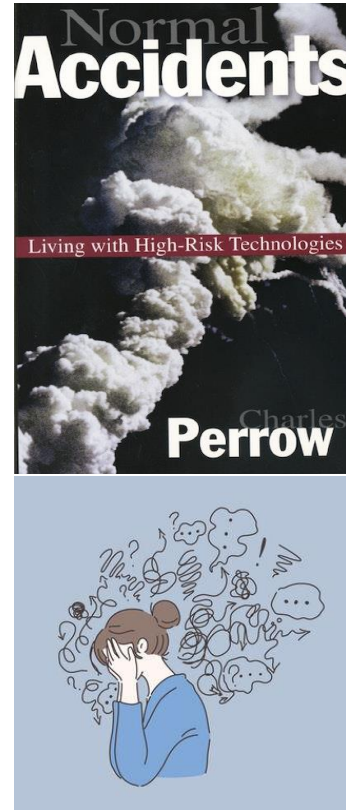
1,086

63.4K



Theoretical background

- Post-industrialized society, complex system, and “normal” accidents (Perrow 1985)
 - An overheated bearing of a train; An equipment failure of a chemical plant
 - This type of technological/complex disaster is not well documented in International Disaster Dataset
- Social media environmental anxiety
- While environmental inequality in physical health has received considerable scholarly attention, few studies have examined environmental inequality in terms of mental health outcomes.



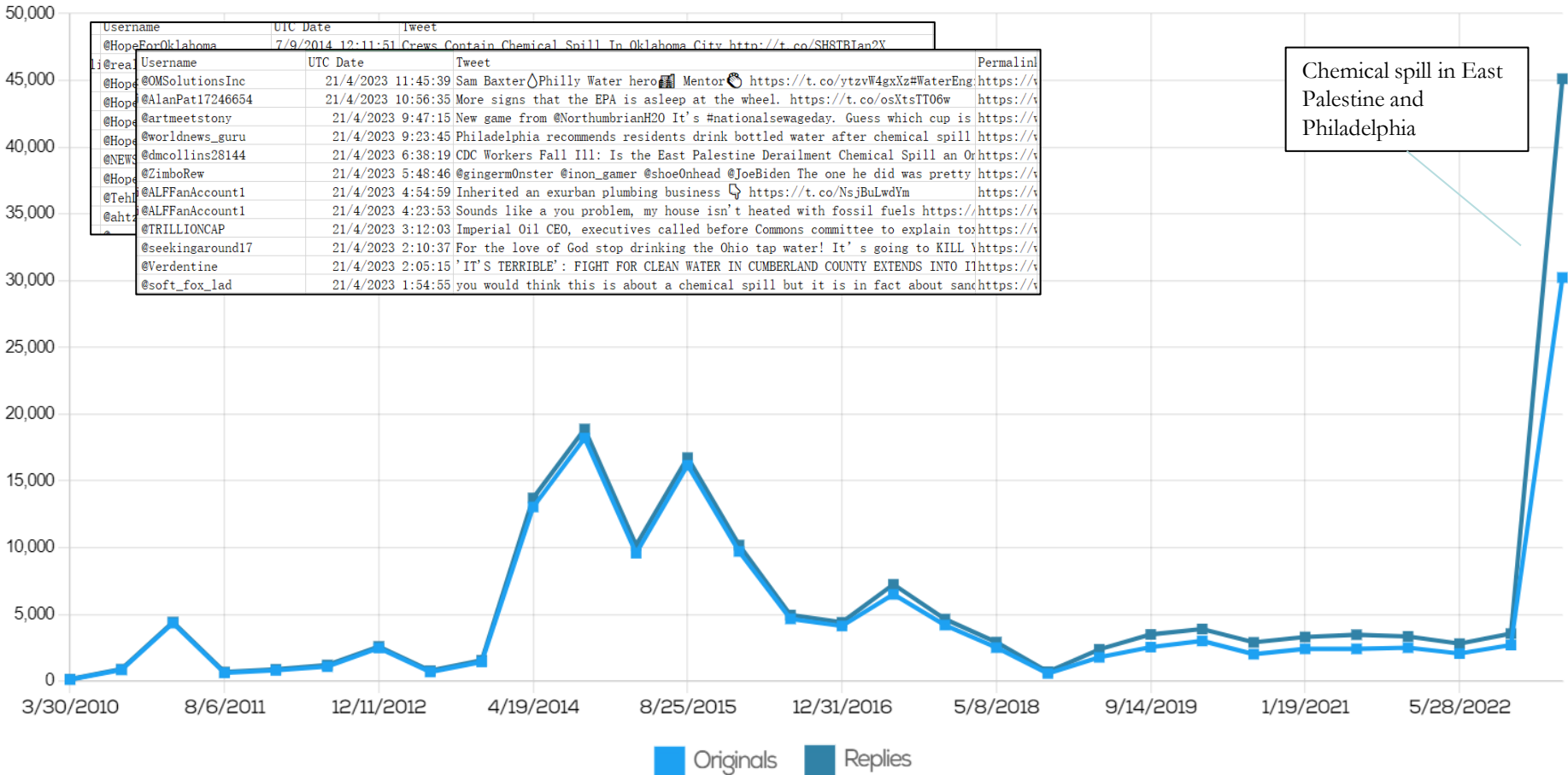
Hypotheses

- **“Where there's smoke there's fire”**: social media environmental anxiety is correlated with preceding environmental accidents.
- **“If men define situations as real, they are real in their consequences” (Thomas theorem)**: environmental anxieties on social media have negative impacts on short-term mental health among people living in the place of incident.
- **“No man is an Island entire of itself”**: There is a spillover of environmental anxieties, and the mental health of those who do not reside at the incident site is also affected.

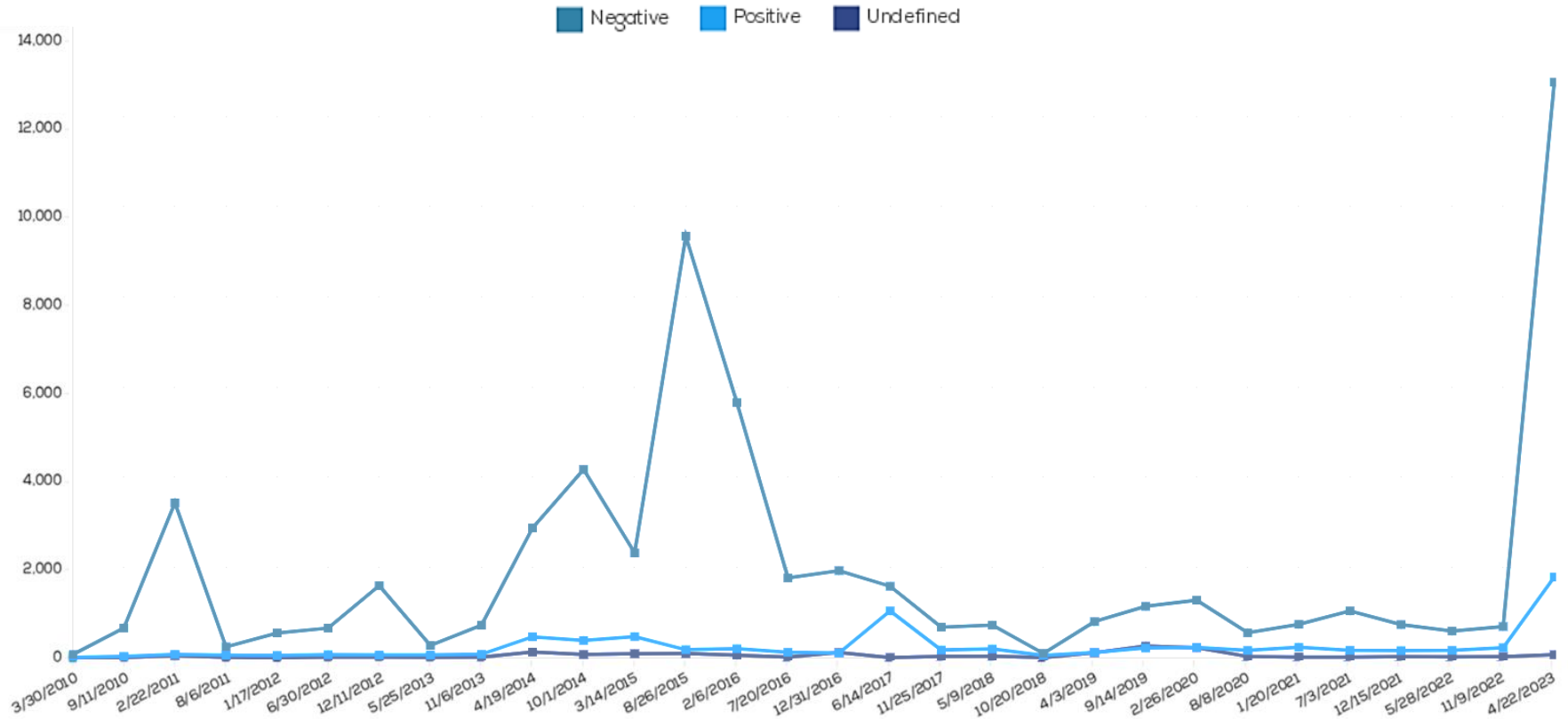
Data

- Twitter data, 2010-2023 (515k)
 - "toxic spill" OR "chemical spill" OR "philly tap water" OR "philly water" OR "illinois tap water" OR "illinois drinking water" OR "indiana tap water" OR "indiana drinking water" OR "michigan tap water" OR "michigan drinking water" OR "ohio tap water" OR "ohio drinking water" OR "pennsylvania tap water" OR "pennsylvania drinking water" OR "wisconsin drinking water" -RT -polo -Chapel -vote -Infrastructure
- Survey data
 - National Health Interview Survey (annually, data released up to 2021)
 - Panel Study of Income Dynamics (PSID) (biennially, data released up to 2019)
 - Restricted geocoded information + date of the interview (criteria for merging)

Count of tweets and retweets over time

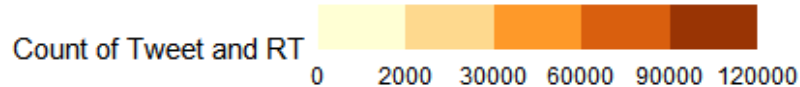
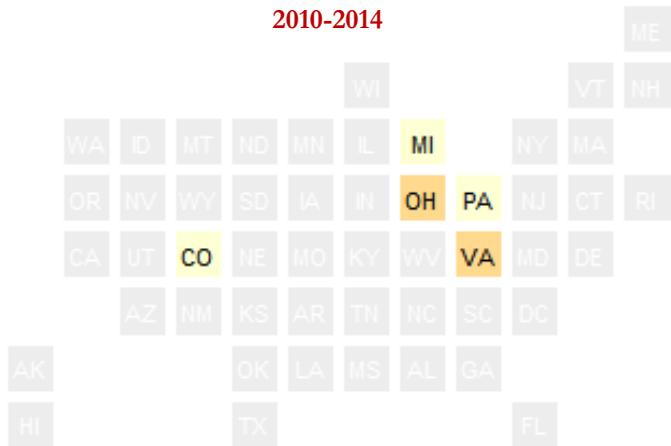


Sentiment of the tweets: predominately negative

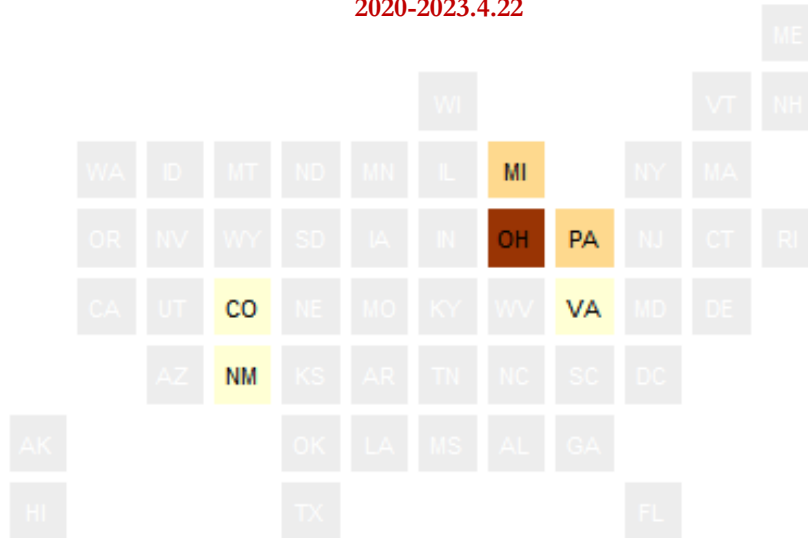


Tweets and re-tweets by period and by location: increased and concentrated

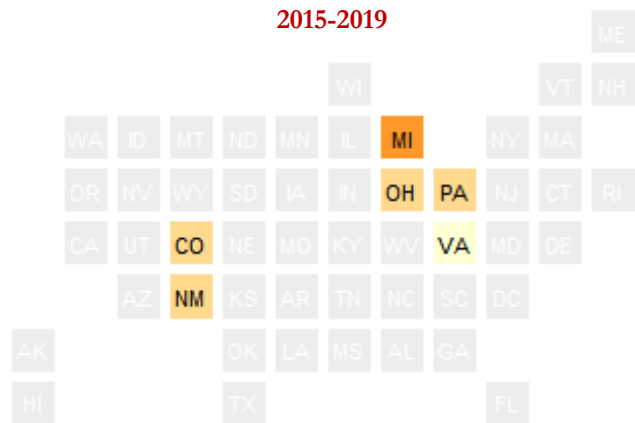
2010-2014



2020-2023.4.22

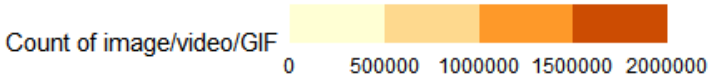
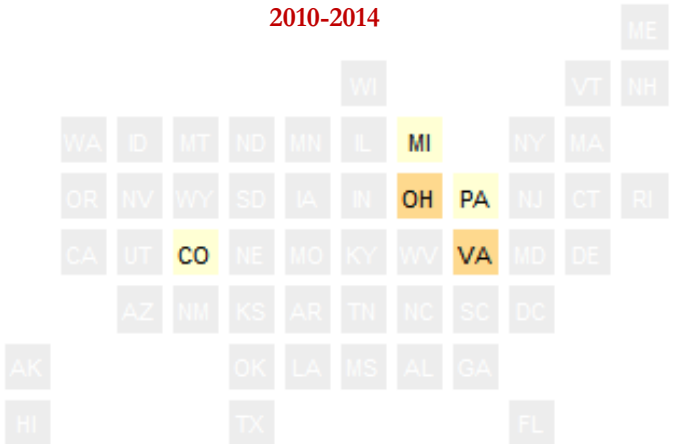


2015-2019

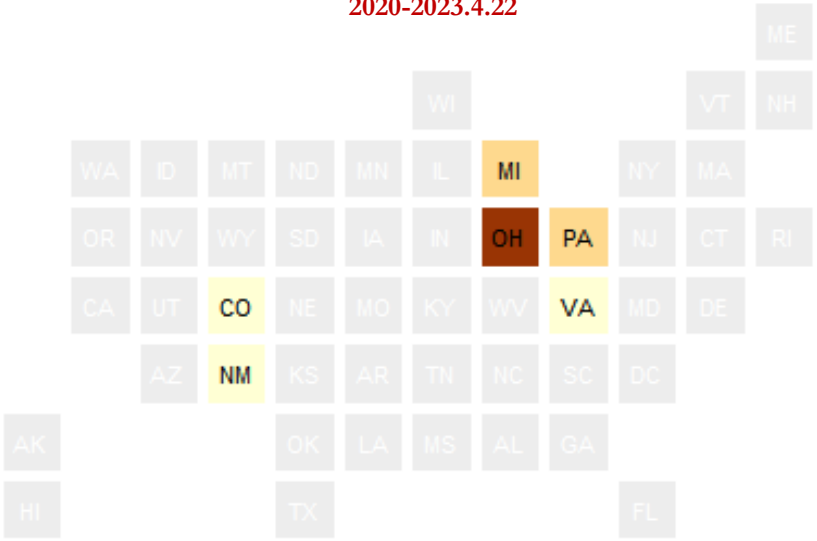


Inclusion of mixed media in Tweets by period and by location

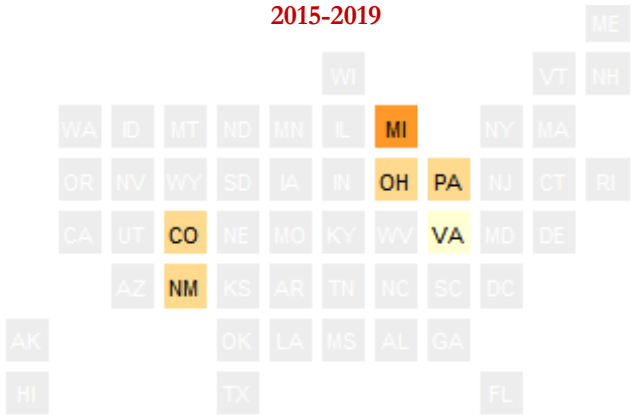
2010-2014



2020-2023.4.22

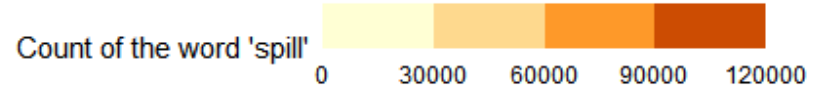
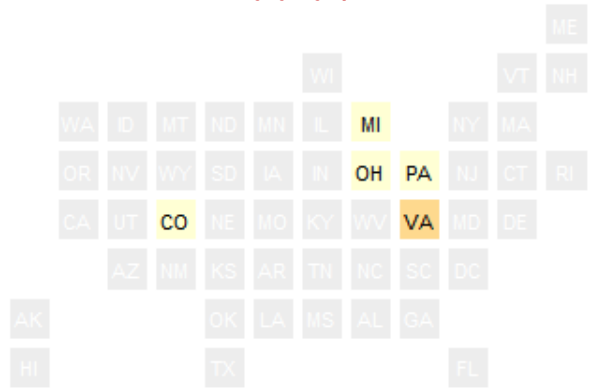


2015-2019

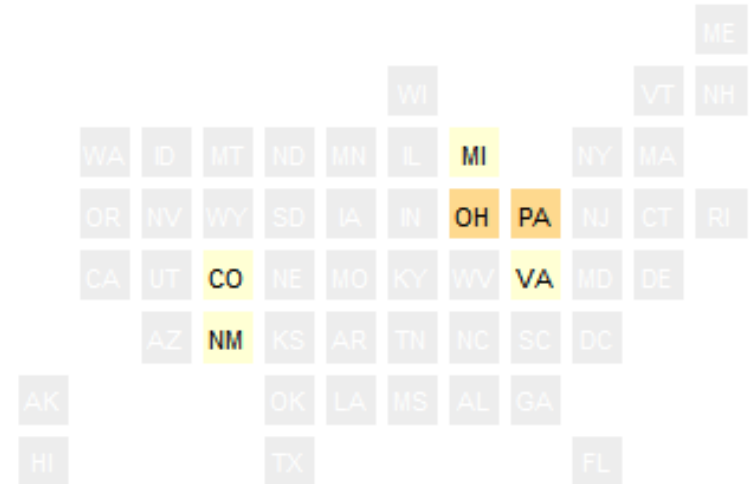


Count of the word “spill” by period and by location

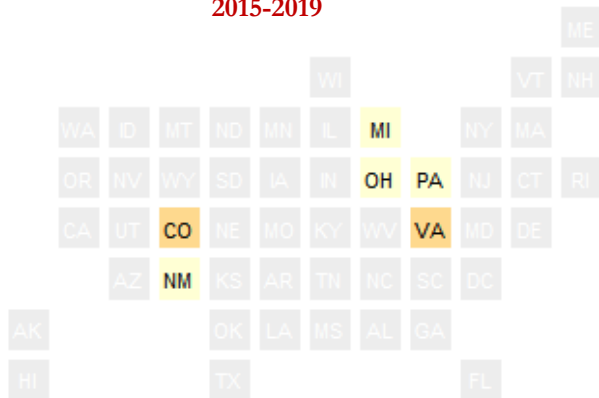
2010-2014



2020-2023.4.22

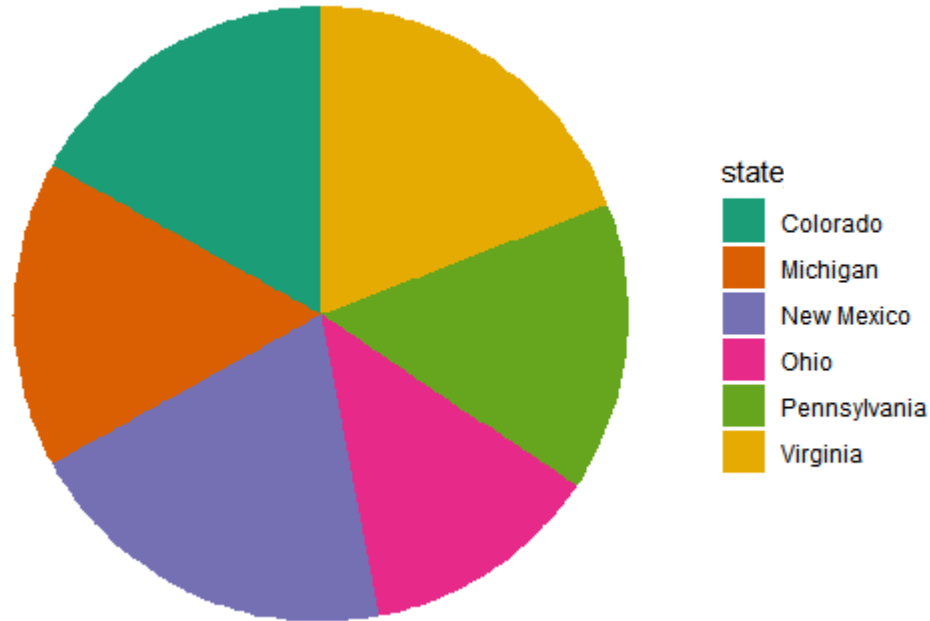


2015-2019



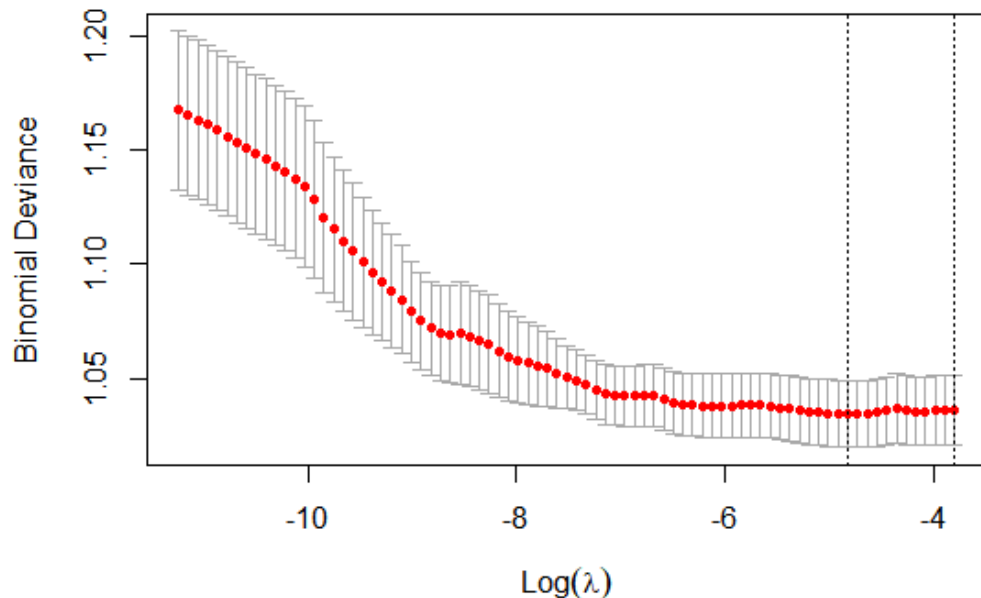
Linking Twitter data to survey data based on location and time

Proportion of mental health decline
(current mental health measure – baseline mental health measure and dichotomized)



Any terms predict decline in mental health?

134 114 100 83 74 61 52 45 36 27 17 11 1



Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	-1.254776	0.059701	-21.018	<2e-16	***
first	-0.157326	0.088328	-1.781	0.0749	.
investig	-0.155348	0.110203	-1.410	0.1586	
just	-0.060535	0.078989	-0.766	0.4435	
live	0.204415	0.129116	1.583	0.1134	
month	0.001177	0.053842	0.022	0.9826	
need	0.105736	0.090169	1.173	0.2409	
ohio	-0.011007	0.004844	-2.272	0.0231	*
peopl	0.006955	0.023868	0.291	0.7707	
poison	-0.008130	0.045178	-0.180	0.8572	
test	-0.014760	0.040838	-0.361	0.7178	
today	-0.004648	0.046573	-0.100	0.9205	
turn	0.079738	0.037400	2.132	0.0330	*

Next steps

- Try linking the two datasets with different lags in time
- Try using more granular survey data
- To use ensemble methods, bagged several prediction from different machine learning methods to get better prediction