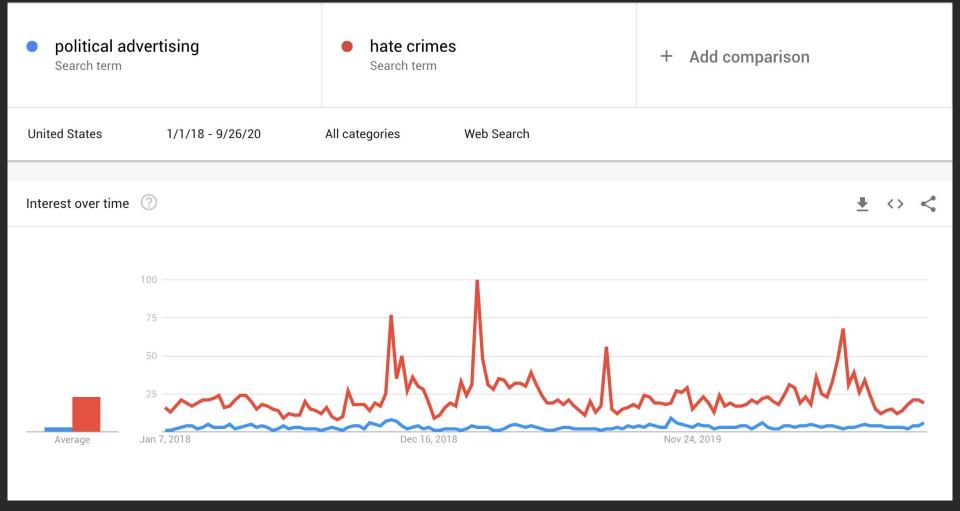


# **Political** Advertising's potential impact on **Hate Crimes** in the U.S

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Designing Data Products Fall 2020









## **Observation**

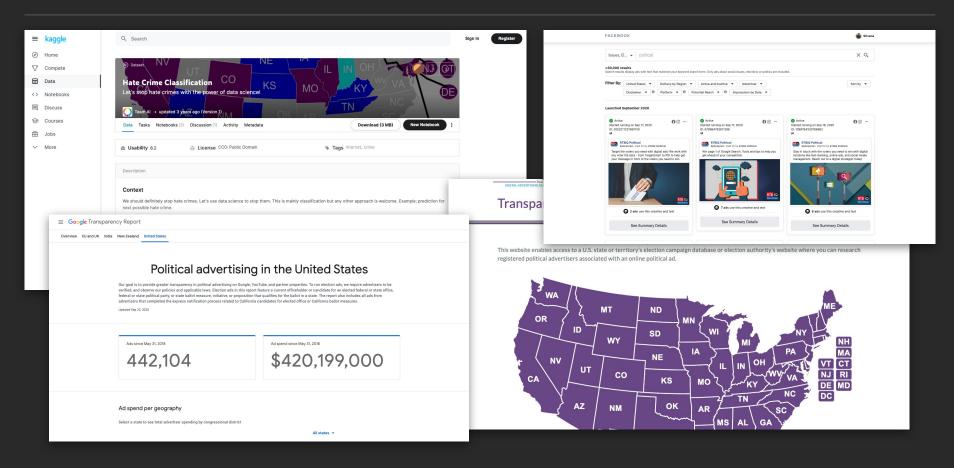
Targeted advertising can influence people's behavior but it is hate crimes that usually create more media attention. We also noticed high volumes of political advertising around election months.

## **Business Question / Hypothesis**

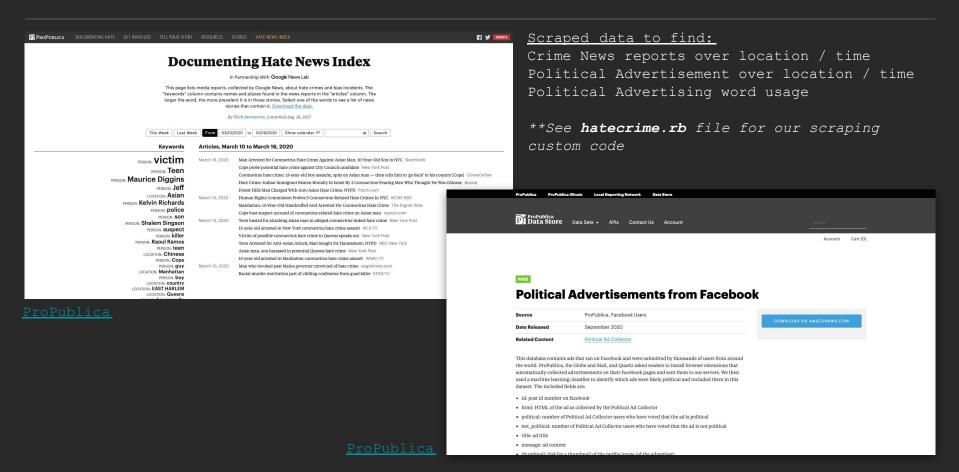
Is there a connection between the *seasonality* of political advertisements in the U.S and the amount of hate crimes reported in that location? Would a supervised model provide any indication of this trend?

From a business perspective, could local and state governments better prepare crisis management planning around certain seasons? Is there enough data available to support a supervised learning approach?

# **Process: Data & Hypothesis Formulation**



# **Data Scraping - Hate Crimes**





Ly US Retirement Bureau Sponsored - Paid for by Senior Financial Services -

FREE Retirement Newscast!

Discover The Little-Known Retirement Savings Secrets!



Protect Your Retirement From Potential Account Freezes

Discover The Closely Guarded Formula



2

- 1: The advertiser's name and disclosure string
- 2: The ad's text, image, and link
- 3: The information Facebook provides about how the ad was targeted
- 4: When the ad was shown to you
- 5: Your browser language

## What We Don't Collect:

Anything personally identifying, including: Your Facebook ID number Your name, birthday, friend list, etc. How you interacted with ads or anything else

#### Why You're Seeing This Ad Only you can see this

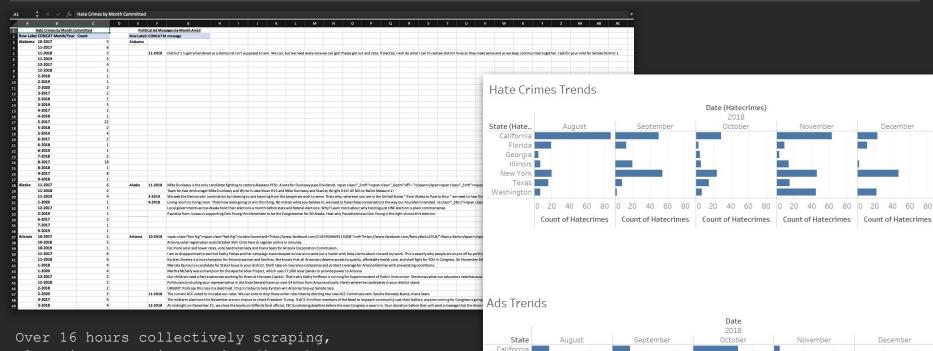




You're seeing this ad because your information matches US Retirement Bureau's advertising requests. There could also be more factors not listed here. Learn More

- US Retirement Bureau wants to reach people who may be similar to their customers.
- US Retirement Bureau is trying to reach people, ages 18 and older.
- US Retirement Bureau is trying to reach people whose primary location is the United States.

# Data Clean Up - Political Ads



Georgia

Count of Message

Count of Message

Count of Message

10

Count of Message

Count of Message

Washington

Wisconsin

Over 16 hours collectively scraping, cleansing, sourcing, and coding the data.

This influenced which models we were able to consider

## **Models**

We did a linear regression to assess whether there are more hate crimes during election months, as well as to identify what the ideal lag time between political ads and a spike in hate crimes using several dummy variables showing different lags before and after elections

Two models tested - by seasonality and state

Shortcomings of the models:

- R<sub>2</sub>: The model explained roughly 25% of the observations
- Adjusted R<sub>2</sub>: .23
- Explanatory Variables not enough statistical relevance (p-values > 0.05%, high likelihood of Type I Error)
- Number of observations: 38 Not enough publicly available data on both crimes and ad data

# Hate Crime Articles	Elec Month -2	Elec Month -1	Elec Month 0	Elec Month +1	
214	0	0	0	0	
236	0	0	0	0	
118	0	0	0	0	
499	0	0	0	0	
462	0	1	0	0	
80	0	0	0	0	
849	0	0	0	0	
666	0	0	1	0	
252	0	0	0	0	
306	0	0	0	0	
317	0	0	0	1	
144	0	0	0	0	
80	0				

744	U	
80	0	
219	0	
258	0	
244	0	
410	0	
237	0	
384	0	
45	0	
359	0	
206	0	
262	0	
1073	0	

**!	see	Ad	and	Crime
Pi	vot	Con	mparı	ison
bу	Mor	nth-	-Yeaı	.xlsx

U	U	U	
State	# Of Politica	l Ads	# of hate crimes
Alabama		1	100
Alaska		6	19
Arizona		18	116
California		30	1510
Colorado		11	152
Connecticut		2	201
Delaware		2	46
Florida		47	476
Georgia		20	180
Hawaii		1	18
Idaho		5	51
Illinois		8	738
Iowa		4	144
Kansas		9	288
Kentuck		1	0
Kentucky		3	205
Louisiana		2	184

# Predicting hate crimes by seasonality

SUMMARY OUTPU	Т							
Regression St	tatistics							
Multiple R	0.23							
R Square	0.05							
Adjusted R Square	-0.10							
Standard Error	280.30							
Observations	38							
ANOVA								
	df	SS	MS	F	Significance F			
Regression	5	139398.67	27879.73	0.35	0.88			
Residual	32	2514125.88	78566.43					
Total	37	2653524.55						
	Coefficients	Standard Error	t Stat	P-value	Lower 95%	Upper 95%	Lower 95.0%	Upper 95.0%
Intercept	344.39	48.79	7.06	0.00	245.00	443.78	245.00	443.78
Elec Month -2	-103.39	284.51	-0.36	0.72	-682.93	476.14	-682.93	476.14
Elec Month -1	117.61	284.51	0.41	0.68	-461.93	697.14	-461.93	697.14
Elec Month 0	321.61	284.51	1.13	0.27	-257.93	901.14	-257.93	901.14
Elec Month +1	-27.39	284.51	-0.10	0.92	-606.93	552.14	-606.93	552.14
Elec Month +2	-108.39	284.51	-0.38	0.71	-687.93	471.14	-687.93	471.14

# Linear Regression: # of Political Ads in state vs. Total Hate Crimes in state

SUMMARY OUTPUT									Ī
Regression St	atistics								H
Multiple R	0.499								
R Square	0.249								
Adjusted R Square	0.233								
Standard Error	271.982								
Observations	49	-							
ANOVA									
	df	SS	MS	F	Significance F				
Regression	1	1154061.343	1154061.343	15.601	0.000				
Residual	47	3476791.188	73974.281						
Total	48	4630852.531							
	Coefficients	Standard Error	t Stat	P-value	Lower 95%	Upper 95%	Lower 95.0%	Upper 95.0%	-
Intercept	125.518	49.519	2.535	0.015	25.898	225.138	25.898	225.138	
# Of Political Ads	12.327	3.121	3.950	0.000	6.049	18.606	6.049	18.606	

# **Findings**

## **RESULTS**

- We were not able to find advertising data that was relevant to the hate crime data we had, it was essentially impossible to cross reference the date and the location of the advertisements with the amount of hate crimes
  - Therefore, a Naive Bayes model was not feasible as we could not define the prior of political advertisements having an influence in the amount or nature of hate crimes

## WHY THIS ISSUE MATTERS IN A BUSINESS CONTEXT

- This could create a digital business model by licensing for advertising platforms (Facebook, Twitter, TikTok) to local governments who have highly partisan residents, could be used for capacity planning of city and local employees (number of officers needed, etc.)

## **OTHER MODEL OPPORTUNITIES**

- With greater data, one could apply an unsupervised learning approach to determine whether words used in political ads are likely to trigger a peak in Hate Crimes news reports

# **Appendix - Data Sources (URLs)**

- Link: Group 6's Data Sets and Source Files
  - Combined Data + Regression Models: Ad and Crime Pivot Comparison by Month-Year.xlsx
  - FB Ad data set post-data wrangling: Ad\_Data post text formatting.xlsx
  - Ads and hate crime dashboard created from raw data: Ads and crime dashboard.twbx
  - Database created from raw FB ad file: Facebookad.db
  - Original raw FB ads data set file: fbpac-ds-end-US.csv
  - Custom Code for running data scraper on hate crime website: hatecrime.rb
  - Data Scraping Output: hatecrimes\_by\_state.xlsx
- Link: original ProPublica Facebook political ads data source
- Link: original ProPublica hate crimes data source