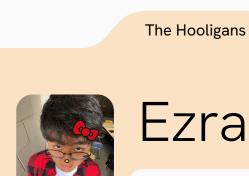
Sentiment Analysis on Articles About the 2024 Election

A. Chyzh, L. Colak, and E. Shin H. Nathan

Don't use our slide show Or else We will sue



Ezra

Rice lover

Watched youtube for the entire program

Sentiment Analysis

Data Analysis



Latif

Baklava lover

Took 5 hour lunch breaks

Background



Siia

Vodka lover

Phone addict

The "Don"

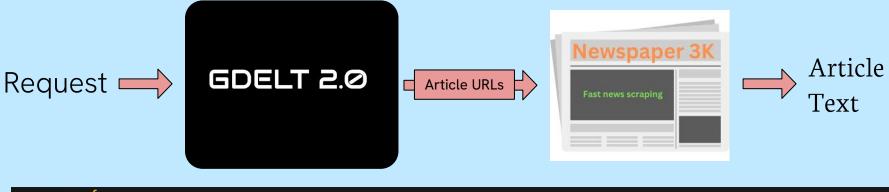


Before U of R grad program



After the U of R grad program

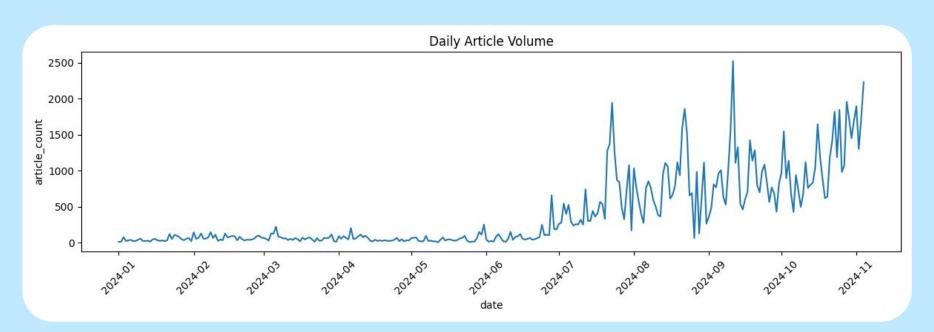
Background & Data

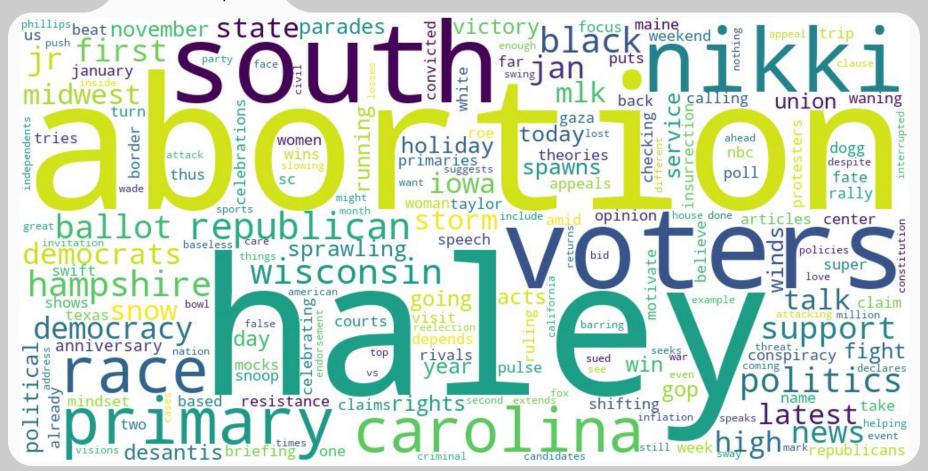


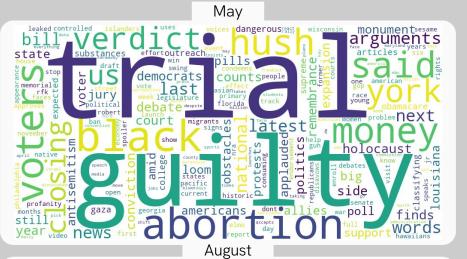
```
params = {
    "query": "harris AND trump AND biden AND sourcelang:english AND theme:ELECTION AND sourcecountry:US",
```

121,689 Articles from between 2024-01-01 to 2024-11-04 From 3225 different news sources

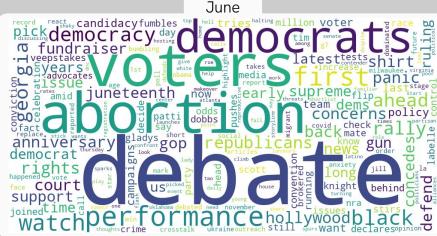
Background & Data



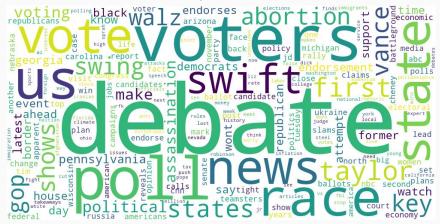




politics winflorida debate watch shows republican mate interview nominee case whitegop case whitegop



September





Polarity: how positive or negative a piece of text is

Averages individual word polarities (word-by-word analysis)

Bias Tagging:

```
df_day['bias'] = "Neutral"

df_day.loc[df_day['more dem'] & (df_day['sentiment'] > 0), 'bias'] = "Pro-Democrat"

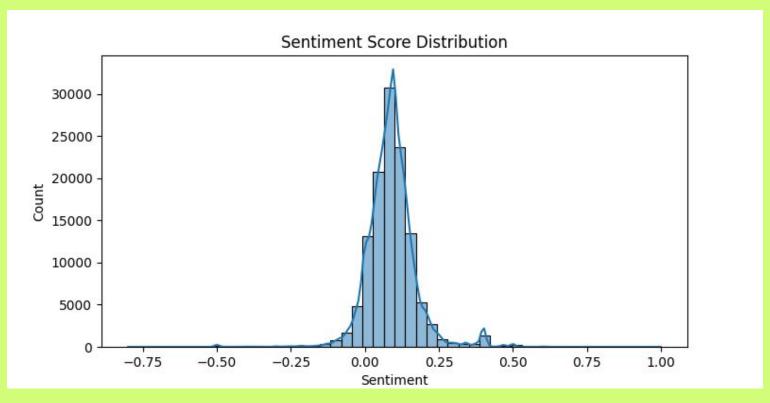
df_day.loc[df_day['more rep'] & (df_day['sentiment'] > 0), 'bias'] = "Pro-Republican"

df_day.loc[df_day['more dem'] & (df_day['sentiment'] < 0), 'bias'] = "Anti-Democrat"

df_day.loc[df_day['more rep'] & (df_day['sentiment'] < 0), 'bias'] = "Anti-Republican"</pre>
```

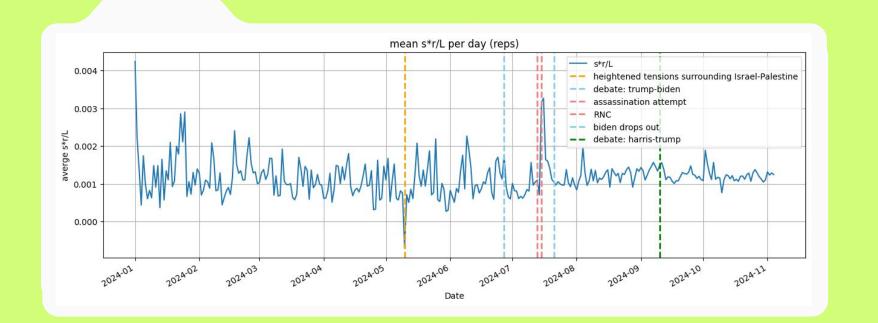
```
republican_keywords = ["Trump", "Vance", "GOP", "Republican", "conservative", "MAGA", "right-wing"]
democrat_keywords = ["Biden", "Harris", "Waltz", "Democrat", "liberal", "progressive", "left-wing"]
```

Sentiment Analysis



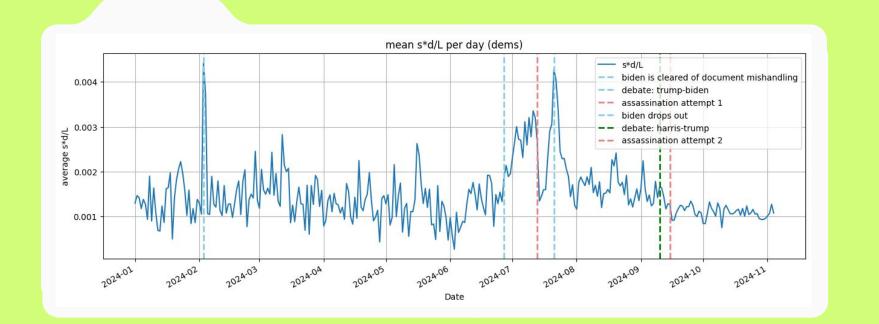
SRDL ScoresTM

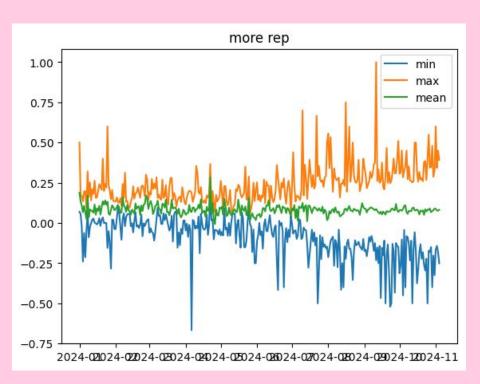
(sentiment)
$$\times \frac{\text{(number of party words in the article)}}{\text{(length of the article)}} = s \times \frac{\text{(r|d)}}{L}$$

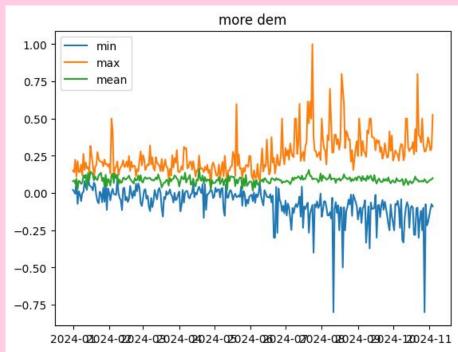


SRDL ScoresTM

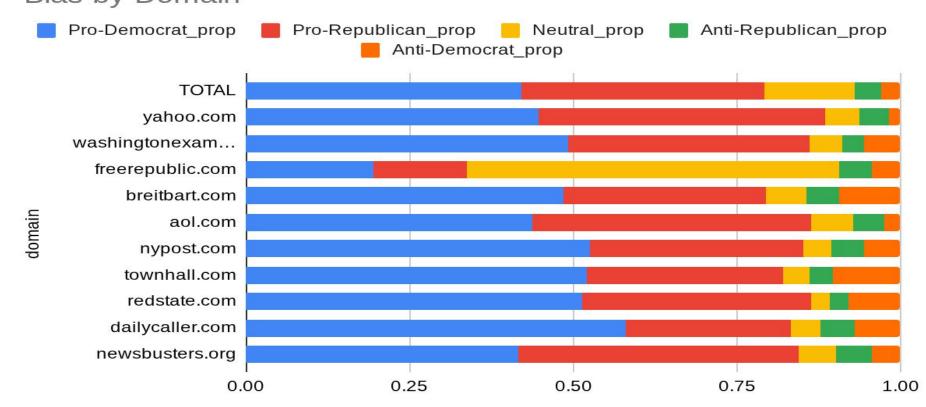
(sentiment)
$$\times \frac{\text{(number of party words in the article)}}{\text{(length of the article)}} = s \times \frac{\text{(r|d)}}{L}$$

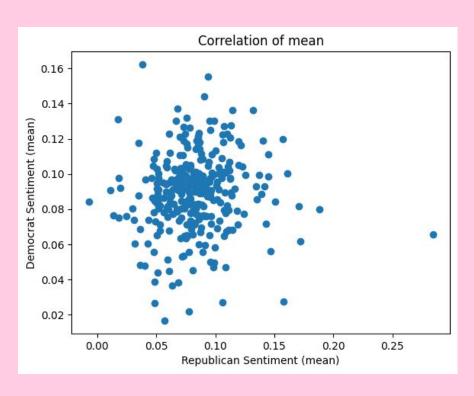


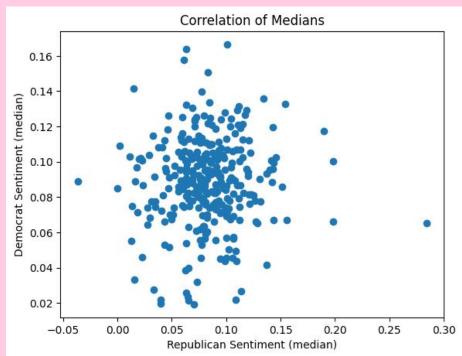






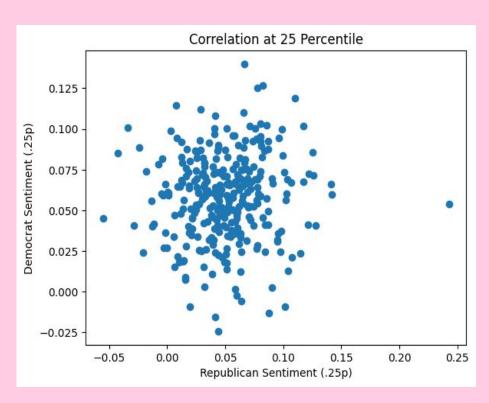


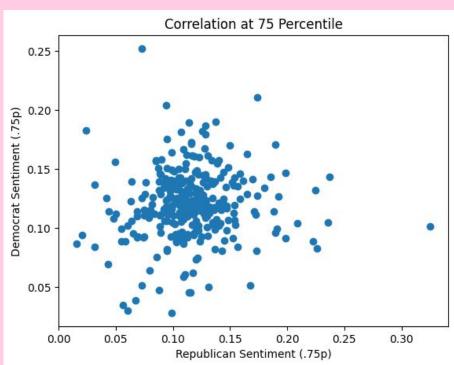




Correlation Coefficient: 0.06

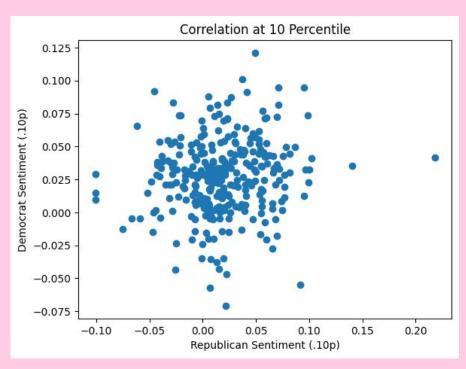
Correlation Coefficient: 0.05

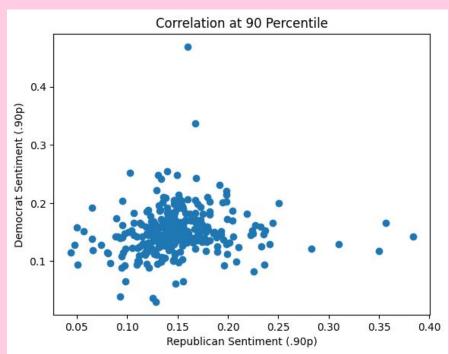




Correlation Coefficient: 0.06

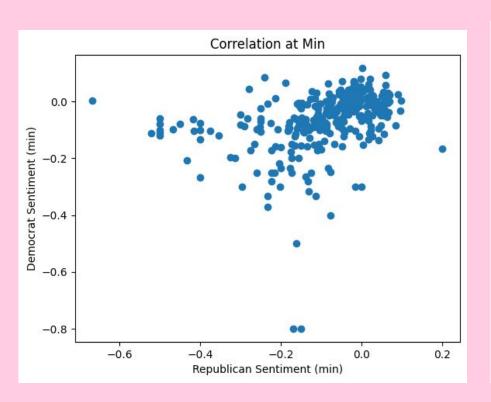
Correlation Coefficient: 0.08

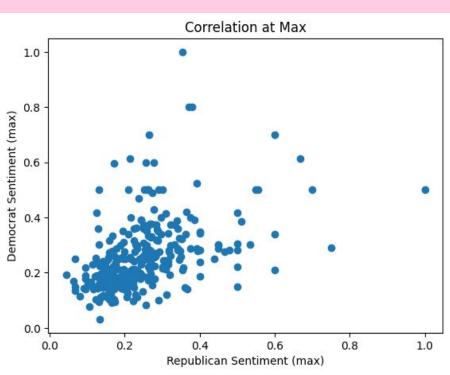




Correlation Coefficient: 0.12

Correlation Coefficient: 0.12





Correlation Coefficient: 0.33

Correlation Coefficient 0.47

Gdelt

Difficult to query

- Slightly mismatched article dates
- Unable to query data before 2017
- Takes a long time

Textblob

Very basic textual analysis

- Limited awareness of context
- No ability to understand sarcasm or irony
- Issues with detecting negative sentiment

SRDL Scores™

Statistically questionable?

- Favors shorter articles
- Unrefined technique for determining which party an article is about
- Uses potentially flawed Textblob sentiment analysis

Us

Python? What's that?

- Dumb, dumber, and dumbest sitting in one room
- Vibe coded our way through this

Conclusion

Our sentiment analysis shows strong alignment with real-world political events.

Moving forward, we will

- expand our dataset,
- calculate additional SRDL scores, and
- explore their potential for predicting election outcomes with greater accuracy.

감사합니 Дякую! Teşekkürler! THANK YOU!!

