

# Analysis of Sentiment around Election Debates

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## Motivation and Objective

## Performance of Candidates

## Evaluate connotation of each candidate's debate topics

## Tweets to try to understand public opinion and draw correlation to debate topics



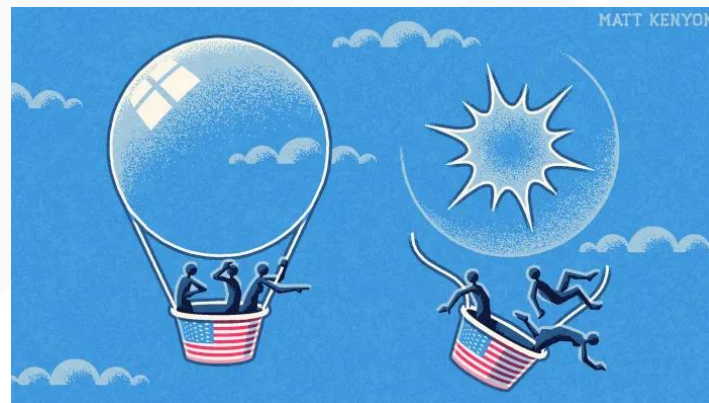
# Dataset

## 1. Debate Transcript

- Debate dialogue
- Speaker
- Timestamps

## 2. Tweets by the General Public

- Used Keywords:  
"Trump", "Biden", "Wallace", "Debate",  
"Election", "Harris", "Tax", "Pence",  
"National Security", "Race in America",  
"Climate Change", "US Economy", "Supreme Court"



# Methodology

## Data Crawling

- Based on snsrape library
- Generate URL with search queries
- Given URL, parse json response from Twitter

## Basic Data Filtering

- Get tweets only written in English
- Exclude short tweets (<20 characters)



# Methodology

## Tokenizing

- Breaking down a corpus to smaller chunks
  - Sentences, phrases or individual words
- Features used to detect the sentiment

## Stopwords

- Words that add little to no extra meaning to the sentence
- EX: a, the, he, she, etc
- Also remove extra punctuation, links, etc

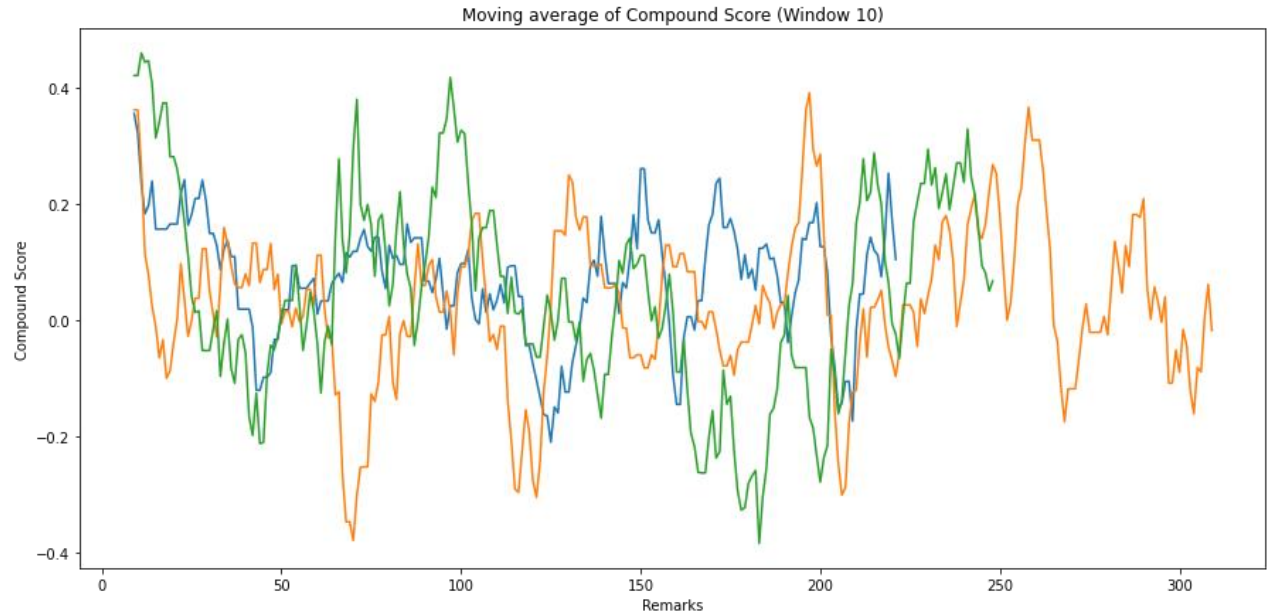
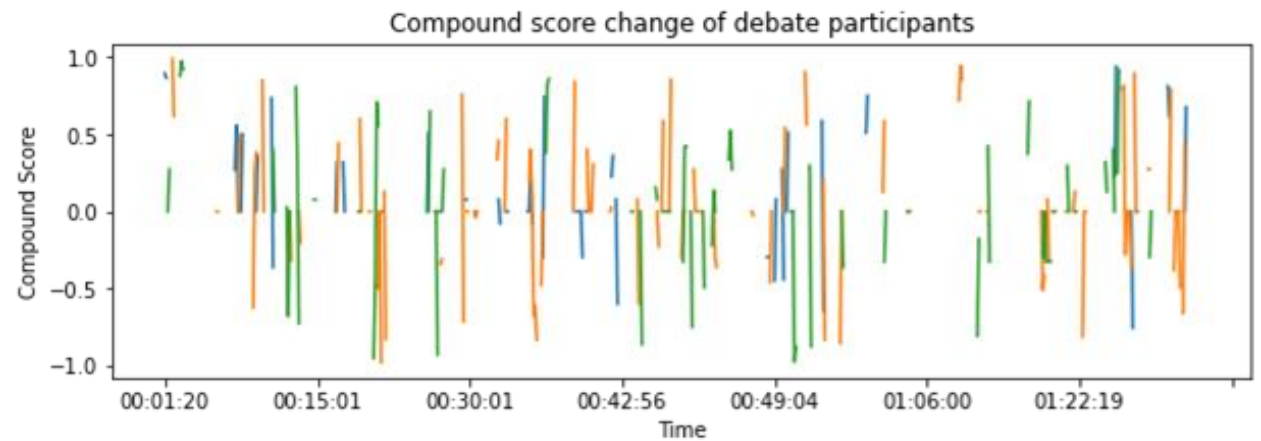
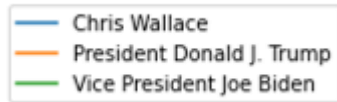


# Transcript Analysis

## Debate 1 [1]

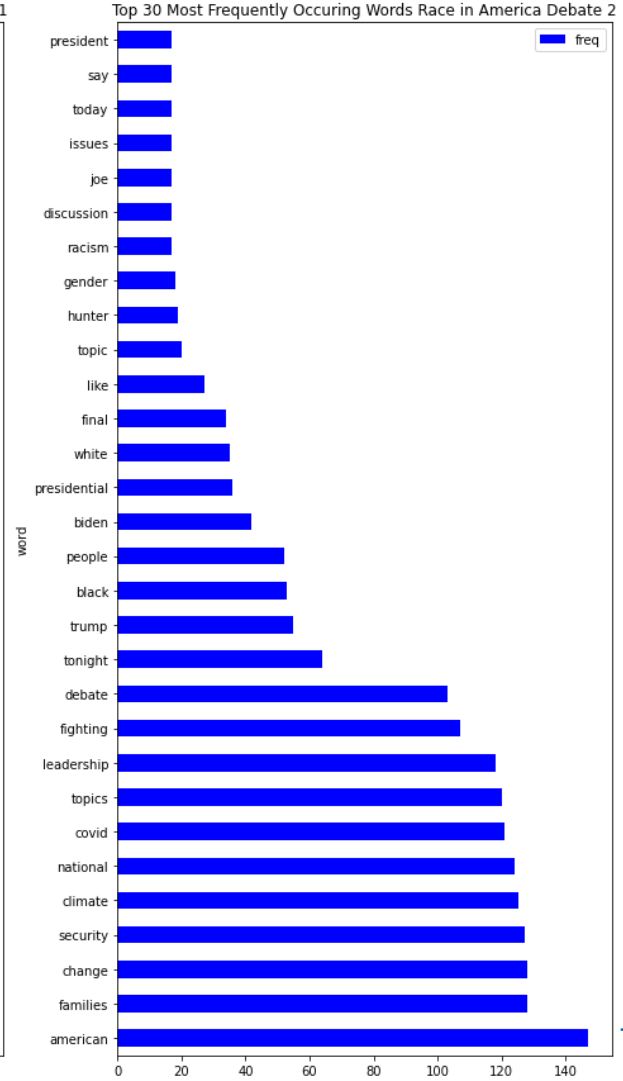
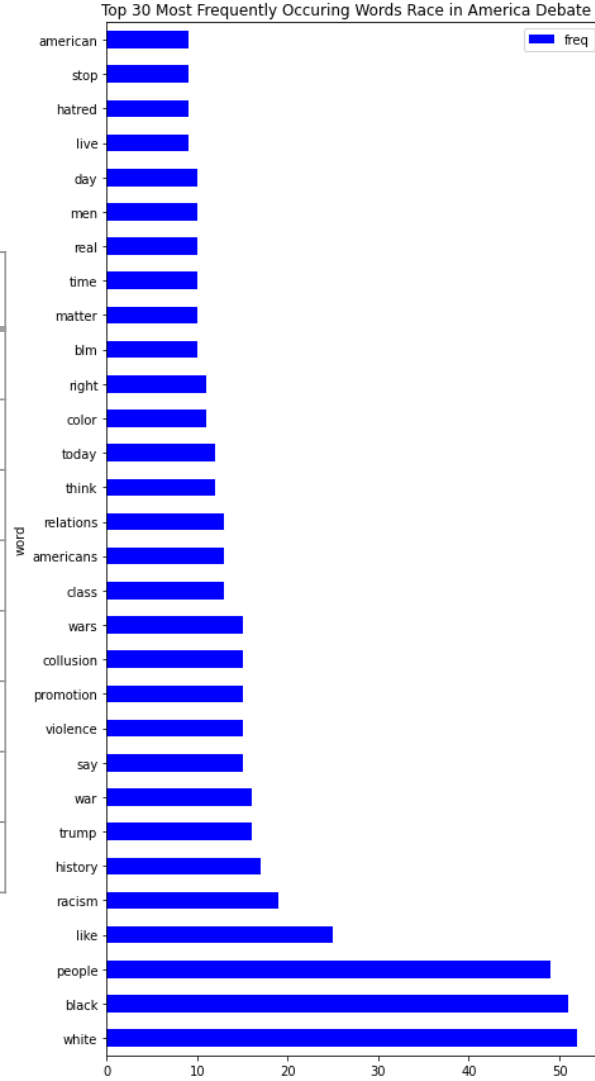
### Words Spoken During Debate 1

- Chris Wallace 4552 words
- President Trump 7234
- President Elect Biden 6620



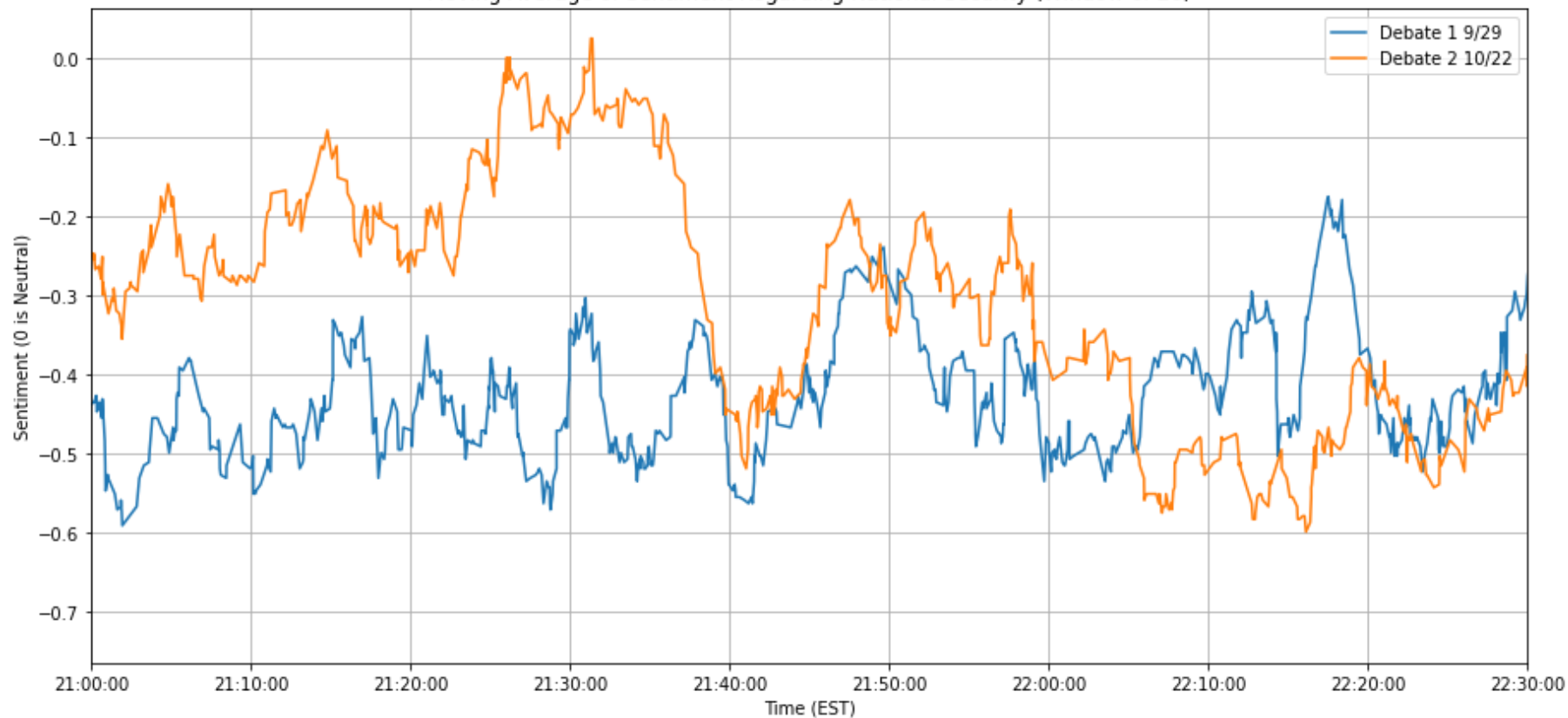
# Race in America

Debate 1		Debate 2	
@torrainewalker	4	@realdonaldtrump	20
@realdonaldtrump	3	@joebiden	10
@joebiden	3	@kwelkernbc	7
@thehill	3	@debates	4
@seanhannity	3	@donaldjtrumpjr	3
@bernlennials	3	@nytimes	3
@townsmarquis	3	@emilywestfahey	3
@rffunke	3	@wftate4	3



# General Opinion: National Security

Moving Average of Sentiment Regarding National Security (Window of 20)

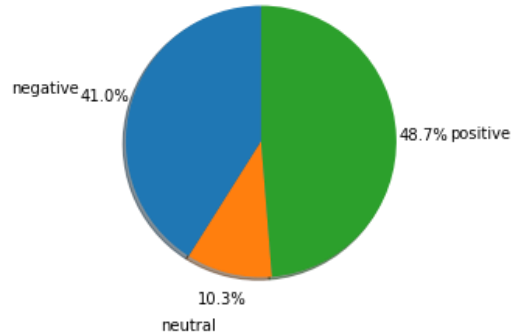




# US Economy

## Debate 2

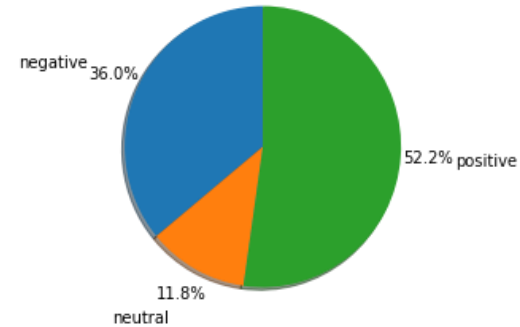
## US Economy



## WordCloud - Comments for Debate1



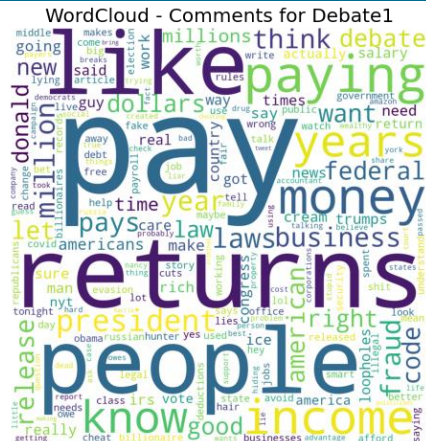
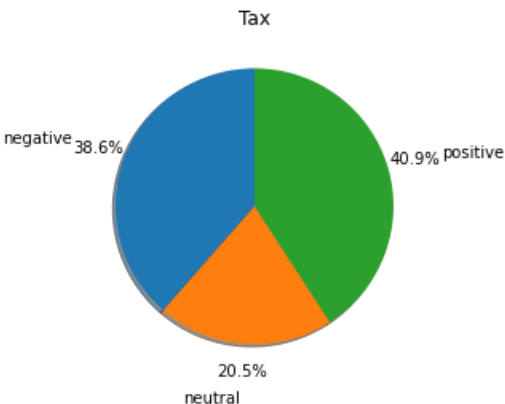
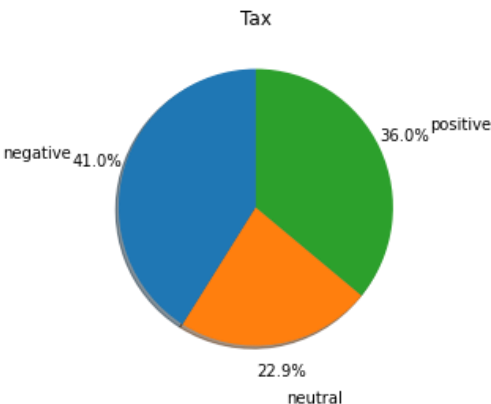
## US Economy



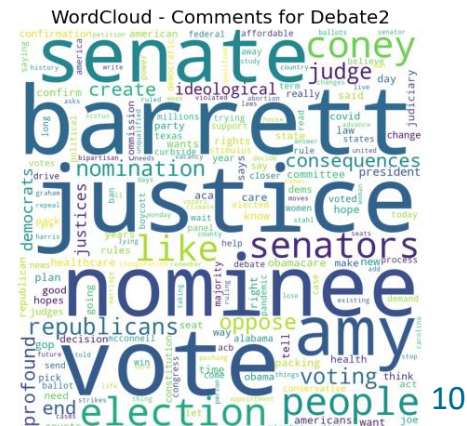
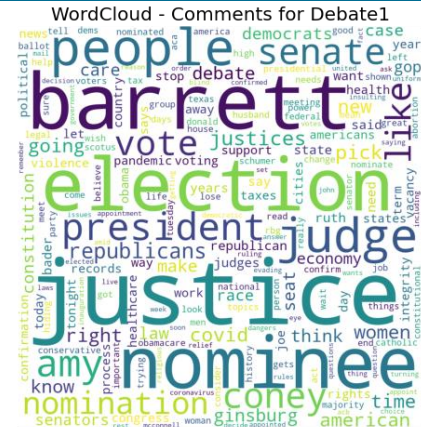
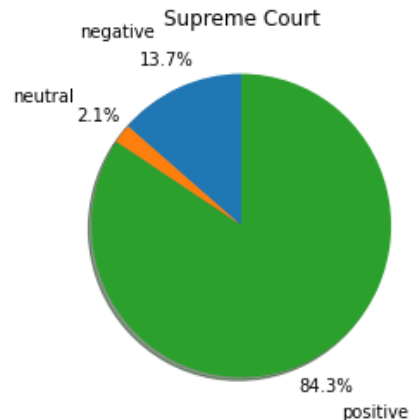
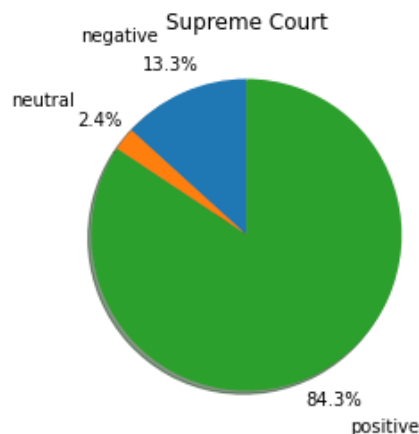
### WordCloud - Comments for Debate2



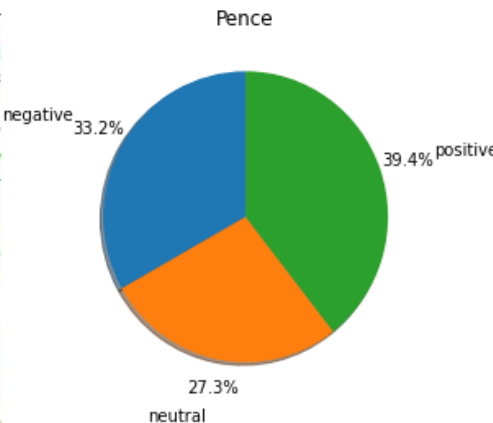
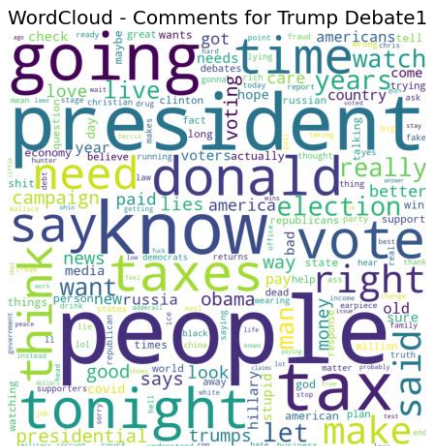
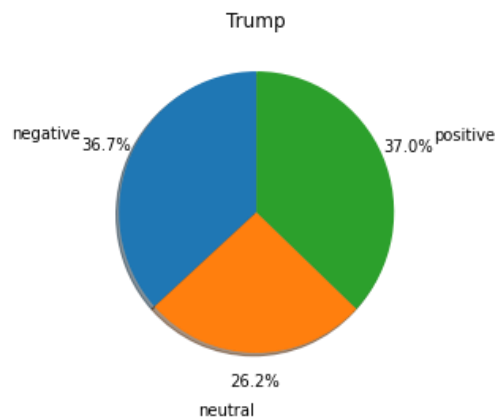
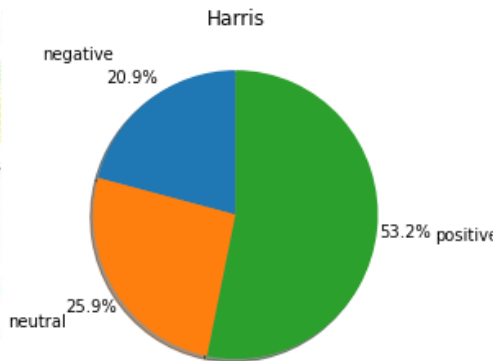
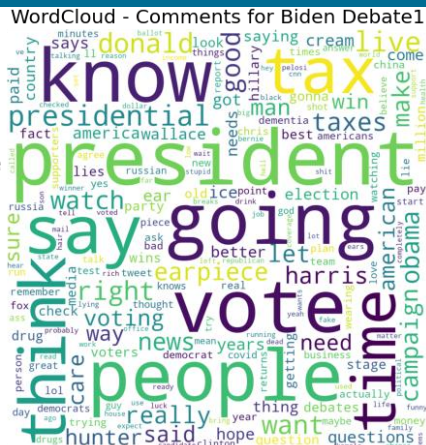
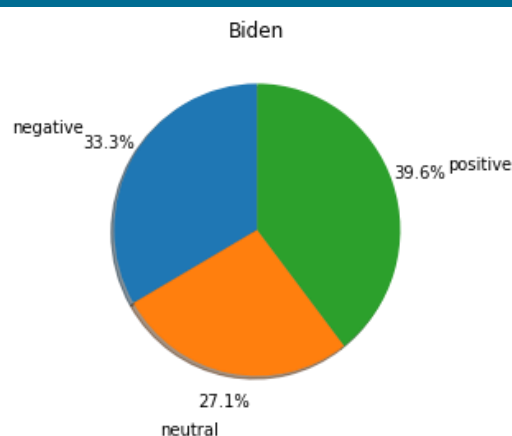
# Taxes



# Supreme Court

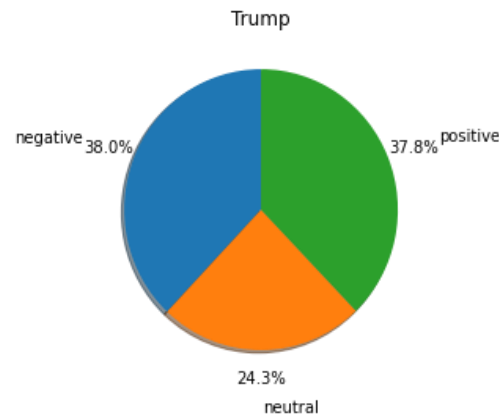
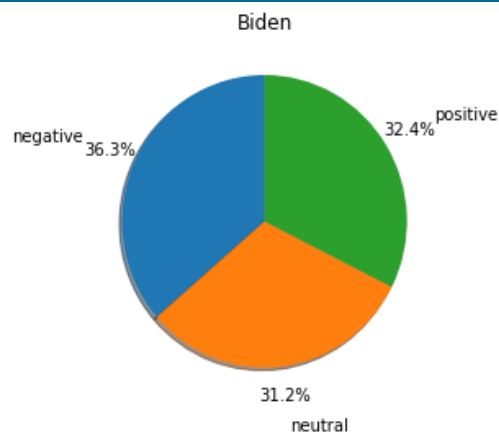


# Running Mates

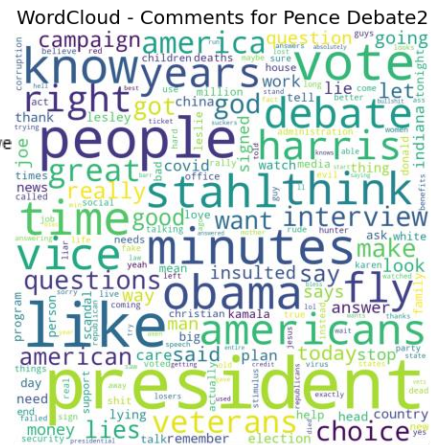
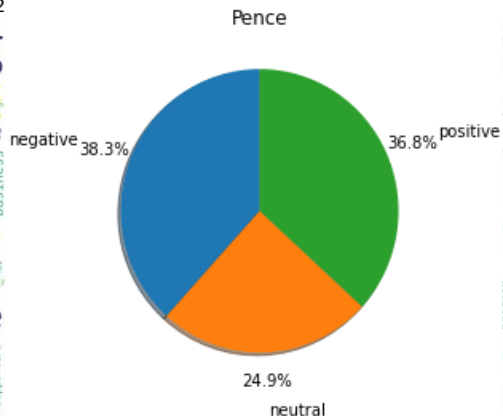
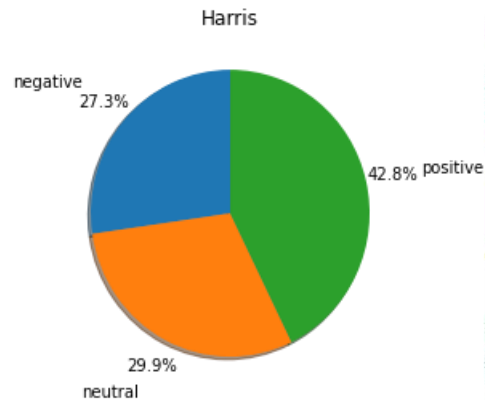




## Debate 2 Candidates

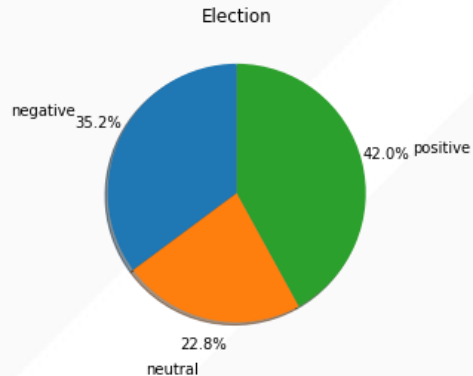
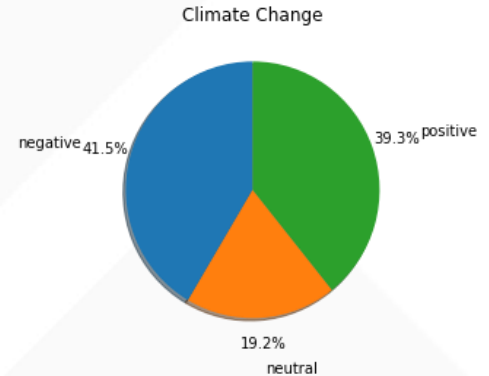
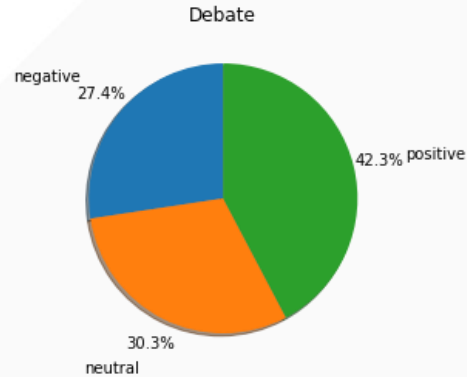


# Running Mates



# Conclusion

- Sentiment Analysis is hard for complex problems
- The actual spoken words during the debate was generally more negative
- Candidates seemed to have been mentioned in all the topics we searched for
- Sentiment was not overwhelmingly indicative of the results of the election



# Citations

[1] Hutto, C.J. & Gilbert, E.E. (2014). VADER: A Parsimonious Rule-based Model for Sentiment Analysis of Social Media Text. Eighth International Conference on Weblogs and Social Media (ICWSM-14). Ann Arbor, MI, June 2014.

# Thank You

Any Questions?

# Prediction of Sentiment

	Biden	Trump	Debate	Election	ClimateChange	Economy	Race	SupremeCourt	Tax	Harris	Pence	Wallace	Walker
Classification													
Random Forest	0.49	0.46	0.51	0.53	0.42	0.60	0.33	0.40	0.50	0.47	0.46	0.56	0.51
Logistic Regression	0.58	0.51	0.46	0.60	0.42	0.42	0.45	0.40	0.54	0.71	0.54	0.58	0.61
Decision Tree	0.48	0.52	0.39	0.55	0.46	0.46	0.43	0.53	0.47	0.46	0.51	0.62	0.47
SVM	0.63	0.62	0.67	0.65	0.60	0.67	0.68	0.69	0.62	0.78	0.65	0.73	0.66
CNN	0.68	0.71	0.75	0.72	0.73	0.76	0.68	0.70	0.72	0.79	0.74	0.73	0.69