

# POLARIZATION AND TOXICITY

DURING THE LATEST
PRESIDENTIAL ELECTION



#### THE IDEA:

- Now more than ever, YouTube is seen not only as a platform where to share and watch videos, but where exchanges between users can be had
- Political discussions and conversations can be seen flourishing under specialized news outlet's videos
- Threads under videos from these news outlets, belonging to the entire political ideology spectrum, can offer us a view on how users belonging to different political leanings interact with each other

### THE QUESTION:

By analyzing such YouTube conversations, is it possible to identify a signicant correlation between the polarization of it and its level of toxicity?

#### WHICH COMMENTS?

 Comments under national, local and independent news channels' videos, together with organizations' channels videos, during the latest US Presidential Elections of 2020, which could be considered as fertile ground for political debates

#### **GETTING THE DATA:**

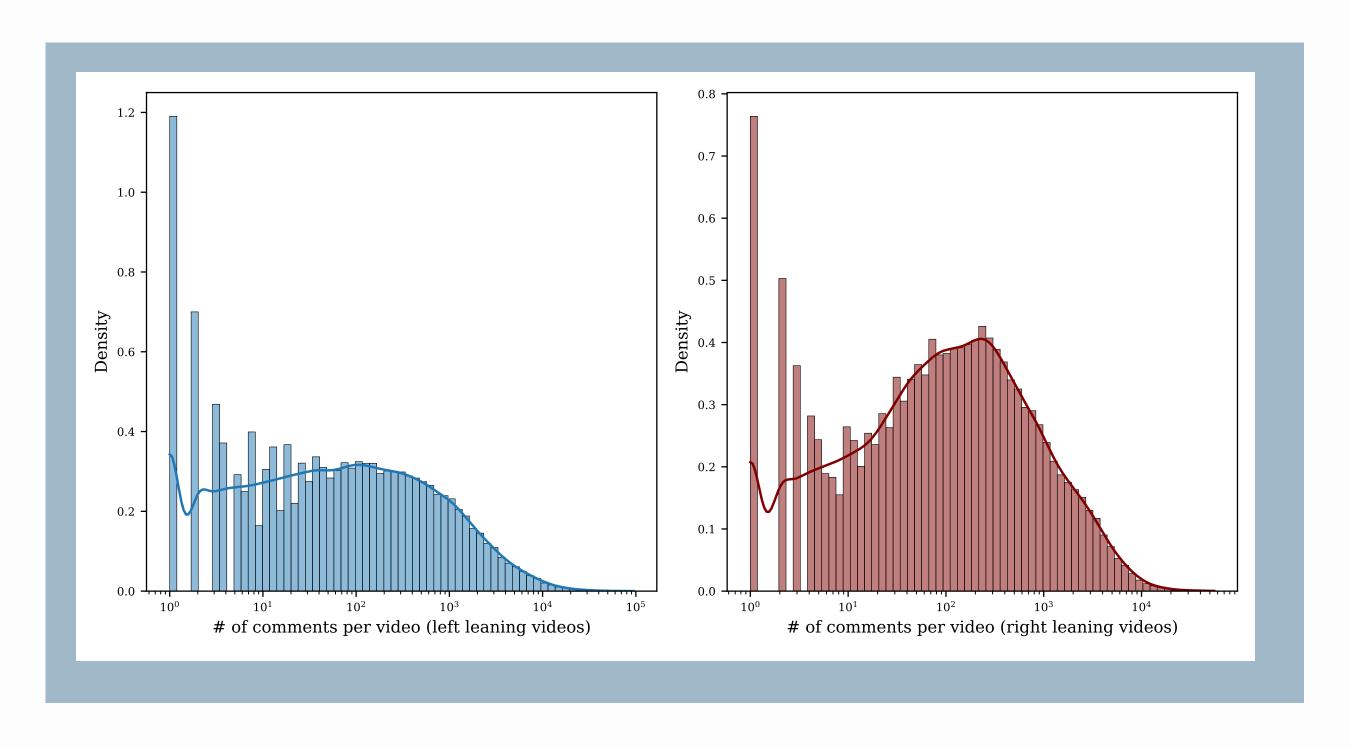
- Almost 1300 channels considered, classified either as left, right or center leaning
- A total 230 thousands videos uploaded between January and August 2020
- Full scrape of each video (root comments + replies)

## RESULTING DATASET:

 166k videos were successfully scraped, retrieving almost 80M comments coming from more than <u>9M</u> <u>unique commenters</u>

	Left leaning	Center leaning	Right leaning	Total:
Channels	427	165	419	1011
Videos	88k	22k	56k	166k
Comments	43M	7M	28M	78M

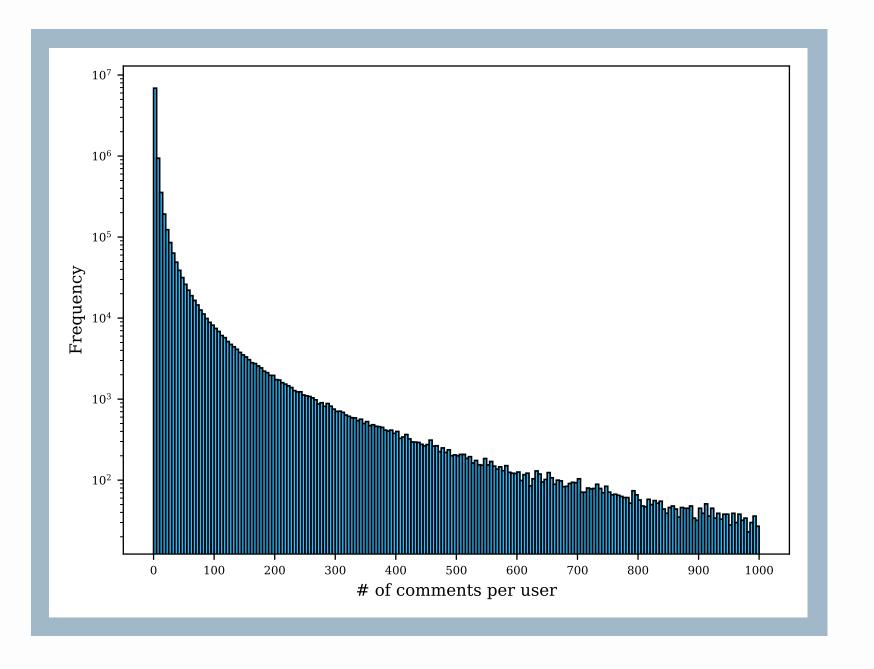
### <u>COMMENTS PER VIDEO:</u>



• Even though there are more videos from left-leaning channels (and thus more total comments), the right-leaning videos have on average more comments per video (mean = 500, median = 91 vs mean = 493, median = 48)

### COMMENTS PER USER:

- 48% of the 9+ million users only have a single comment throughout every video
- "Active" users (those with at least
   5 comments) are 24% of the total



 While the average number of comments per user is 8.7 (median = 2), the distribution decreases in an exponential fashion

## MODELING STEPS:

#### Labeling active users' leaning:

 Averaging the number of comments under left, right and center leaning channels

#### Scoring comments' toxicity:

• Google's Perspective API

# Measuring toxicity of a thread of comments:

 (avg. of the ten percent most toxic comments in thread) - (avg. toxicity in thread)

## MAPPING POLARIZATION:

- We can consider polarization of the users' leaning in a given thread of comments as the extent of users with opposed leanings taking part in it
- One possible way of quantifying it is checking whether the leanings' distribution deviates from a unimodal (unpolarized) distribution and resembles a bimodal (polarized) one
- Many measures and statistical tests are designed to check the bimodality of a distribution, each with its pros and shortcomings: we shall review three of them to find the best suited one for this analysis

## MEASURES OF POLARIZATION:

# Bimodality coefficient:

$$BC = \frac{m_3^2 + 1}{m_4 + 3\frac{(n-1)^2}{(n-2)(n-3)}}$$

 BC ≈ 0.55 for a uniform distribution. Higher values indicate bimodality

# Distance from unimodality:

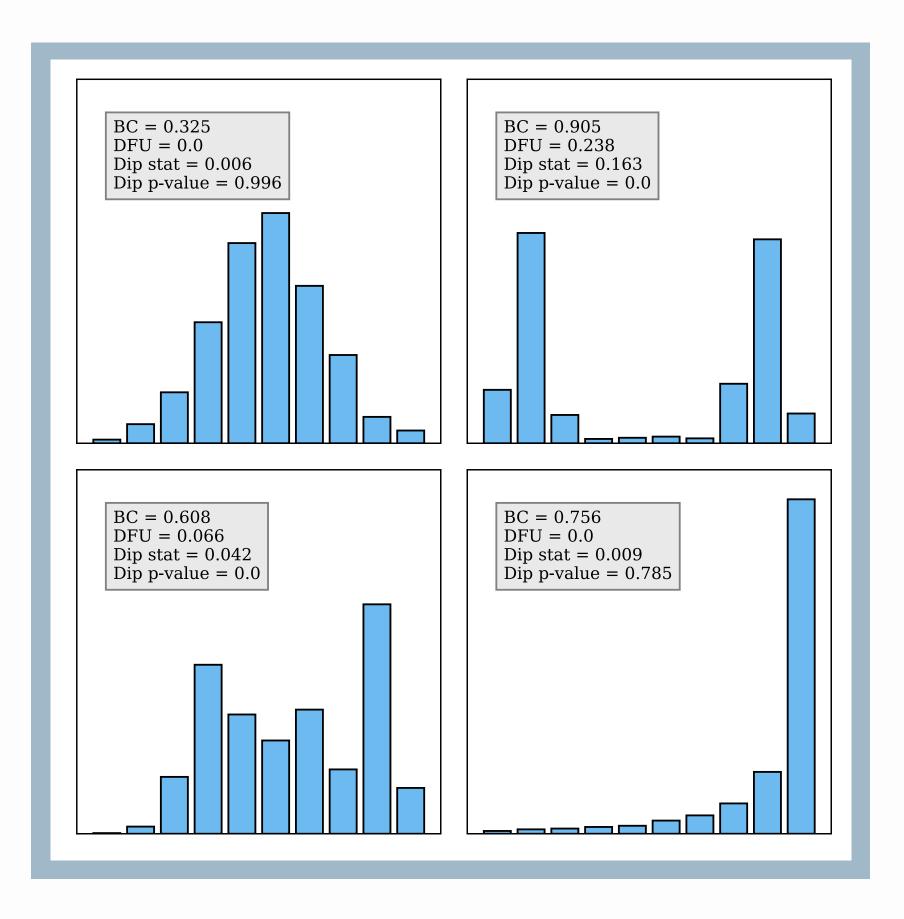
- Bins the data into K ordinal categories, and computes the deviation from the unimodality rule
- Any value higher
   than DFU = 0
   indicates a deviation
   from unimodality

# MEASURES OF POLARIZATION (CONT.):

#### Hartigan's dip test:

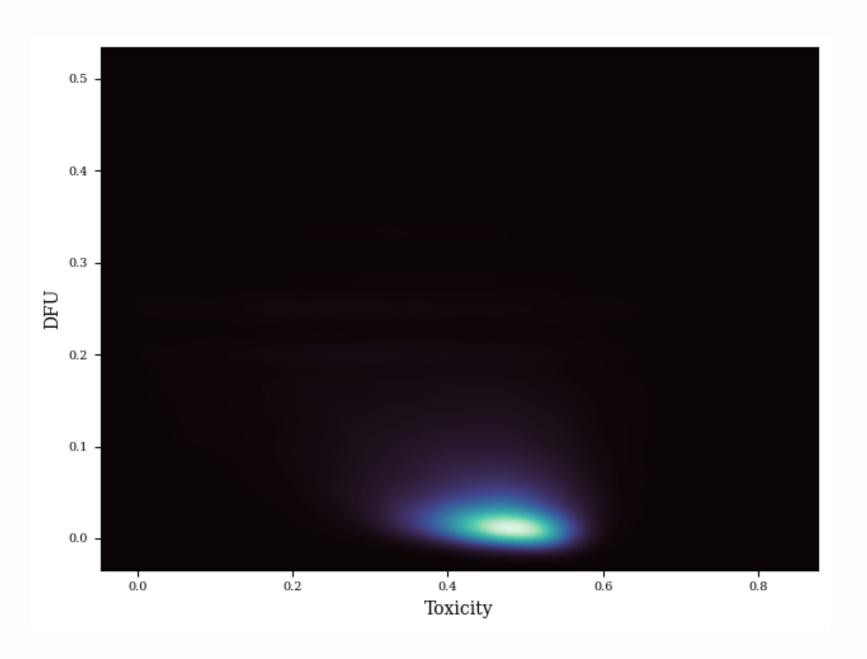
- Measures multimodality in a sample by the maximum difference, over all sample points, between the empirical distribution function, and the unimodal distribution function that minimizes that maximum difference
- The dip test index is higher when the distribution approaches bimodality
- For the dip test p-value, "values less than 0.05 indicate significant bimodality and values greater than 0.05 but less than 0.10 suggest bimodality with marginal significance"

#### COMPARISON:



- In the first panel, the three measures agree on not considering the normal distribution as bimodal
- In the second and third one, the distributions are correctly classified as bimodal, even though the dip p-value doesn't give us any further insight on the extent of the bimodality
- In the fourth one, the BC classifies the skewed unimodal distribution as more polarized than the one in the third panel

## POLARIZATION VS TOXITICY:



- Using DFU with 8 bins to compute the polarization score of a video, there seems to be a slight negative correlation between the toxicity and the polarization of the comments under a video
- It's important to notice how, while almost 30% of the videos have a DFU score = 0, the majority of videos with high polarization are those with relatively few comments
- The average value for the DFU scores is 0.05, the average value for the toxicity scores is 0.43

## POLARIZATION VS TOXITICY:

- Furthermore, there's no appreciable difference between the average toxicity in the comments of left, right or center leaning videos, nor the correlation between polarization and toxicity
- Videos coming from center leaning channels however are, on average, more polarized than those coming from left or right leaning channels

### CONVERSATIONS:

- By considering single conversations under the videos (composed by a root comment + its replies), we can notice how they are on average more polarized (0.12), but with a lower average toxicity score (0.32)
- The correlation between polarization and toxicity is less marked than when considering the whole corpus of comments