Voting Mechanism and Consensus Involving

Shirui Zhou, Sharon Chuang, etc

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Lets Do a Experimentation Before We Start

Polis

- Link: https://pol.is/2hkcy8chhj
- ► Topic: China Zero Covid Protests
- User can agree / disagree on different comments
 - Its anonymous!!.
 - Based in the US
 - ► → youwon' tblacklist!!



Outline

- Motivation
- Research Questions
- Data
- Methods
- Results thus far
- Limitations/Next Steps

Motivation

Problems

- 1.'Horse race' problem and misinterpretation in polls and probabilistic modeling
- 2. Echo chamber and polarization
- 2. No alternative voice in some authoritarian countries
- Solution
 - Voting mechanism served as decentralized opinion aggregation platform
 - Vote on the "comments" by participants
 - Review and understand the most trivial concern of the other side

Prior Research

Polis

- a real-time system for gathering, analyzing and understanding what large groups of people think in their own words, enabled by advanced statistics and machine learning.
- method:
 - 1. Read about the polis opinion matrix
 - 2. Dimensionality reduction
 - 3. Clustering

p.

China Zero Covid Protests

"In memory of the people who died in the fire in Uturnual in November 24th, and the ones who are silenced, under arrest, and have their freedom taken away by the Chinese government with their brutal measure against coxid=19 periodemic.

Understand the Protests in China

- The Toil of 'Zero Covid': The protests against China's strict pandemic policy come after President Xi Jinping's unbending approach hurt businesses and strangled growth.
- At a Tipping Point': For the protesters, public dissent was unimaginable until days ago. Our columnist asked young people what led them to take the risk.
- The Economic Fallout: The growing unrest in the world's biggest manufacturing nation is injecting a new
 element of uncertainty and instability into the global economy.
- Reasserting Control: The Communist Party is drawing on its decades-old policy of repression and surveillance — along with some new tactics — to quash the protests. (NewYork Times, 2022)

Welcome to a new kind of conversation - vote on other people's statements.

Anonymous wrote: In my opinion, it's bravy people, but i don't to b	s remaining oncern of	
⊘ Agree	Ø Disagree	Pass / Unsure

Are your perspectives or experiences missing from the conversation? If so, add them in the box below.

What makes a good statement?

- Stand alone idea
- · Raise new perspectives, experiences or issues
- Clear & concise (limited to 140 characters)

Please remember, statements are displayed randomly and you are not replying directly to other participants' statements.

Share your perspective...

Research Questions

- Observation
 - Static: Opinion landscape of the participants and representativeness of comments
 - Dynamic: Is a consensus forming in the voting?
- Activism
 - Would prioritizing the comments from the opposite 'echo chamber' make a difference?

Data Sources

Vtaiwan open source data

- Experimentation of voting mechanism (45s test)
- ▶ Uber Issue: Should Uber be regulated in Taiwan
- Main data set: participants vote csv, comment csv

Method: Data Acquisition

participants-votes.csv

- meta-data: participant, group-id, n-comments, n-notes, n-disagree, n-agree
- sparse matrix: participants (x-axis) vote on each comments (y-axis)

comments.csv

 variables: comment-id, author-id, moderated, comment-body, timestamp

Method: Process flow

- 1. Data Clean
- 2. Dimensionality reduction:
 - PCA
 - UMAP
- 3. Cluster:
 - Kmean

Method: Data Cleaning

```
1 def count_finite(row):
     finite = np.isfinite(row[val_fields])
2
3 return sum(finite) # count number of True values in
    finite `
4
5 def select_rows(df, threshold):
6 number_of_votes = df.apply(count_finite, axis=1)
valid = number_of_votes >= threshold
8 return df[valid]
9 df_votes = select_rows(df, 7)
11 ## remove statements (columns) which were moderated out
12 statements_all_in = sorted(list(df_comments.loc[
    df_comments ["moderated"] > 0].index.array), key = int
```

Methods: Algorithms



Methods: Dimension reduction

- enables us to visualize participants in relation to each other within the opinion landscape
- Participants are closer together in this landscape if they tend to agree. And further apart if they tend to disagree.



Abbildung 1: Visualize w/ PCA

Abbildung 2: Visualize w/ UMAP

Methods: Dimension reduction



Abbildung 3: Source: Packt_Pub, via Hackernoon

- PCA
 - effective for visualizing clusters or groups of data points and their relative proximities.
 - Identifying the hyperplane which lies closest to the data and then

Methods: Dimension reduction



UMAP (Uniform Manifold Approximation and Projection)

- effective for visualizing clusters of data points.
- nonlinear dimensionality reduction method
- Scalability: can be applied directly to sparse matrices

Methods: K-means

Algorithm

- Randomly initialize k cluster means (here: k = 2)
- 2. Iterate:
- 3. Assign each object to the nearest cluster mean
- 4. Recompute cluster means
- 5. Stop when clustering converges

def	<pre>kmean_get_grouped(embeds, k): grouped_embed = KMeans(n_clusters=2, : grouped_embed = grouped_embed.fit(embel labs = grouped_embed.labels_</pre>										
	A_key = np.where(labs == 0)[0] B_key = np.where(labs == 1)[0] return A_key, B_key										

Describing the analytic sample

Name	Estimand
Dimensions of pre-processed matrix	(1921, 203)
Dimensions of post-processed matrix	(1269, 198)
Total number of possible votes	251262
Total number of agrees	30237
Total number of disagrees	11661
Total without vote	208097
Percent sparse	0.8282072100039003

Full participants * comments matrix

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- 0.00

- -0.25

- -0.50

-0.75

- -1.00

Results: Divisiveness of Comments

- How divisive was the conversation?
- Statement close to zero are voted on the same way - either both opinion groups agree or disagree
- Statement far from zero are divisive participant were split between agreement and disagreement



Results: Consensus Comments

coment	b_support_share	b_disagree	b_agree	a_support_share	a_disagree	a_agree	author- id	comment- id	
我覺得要提前考慮電腦自動駕駛的情况,包括現代大眾系統是否還有興建必要,以及以後計程準									107
我覺得繳稅是一個企業在台灣經營的義務,Uber或其新創模式再麼優秀,在台灣的所得都應該對	0.787611			0.784314					165
應修改讓自用車的乘客保險也能保障乘客的權益或由UBER統									95
我覺得 UberX 目前無法幫乘客保意外強・讓我感到沒有									49
我覺得大眾運輸工具普及後,營業車的空車率已不斷攀升,開放自用車載客不會擴大需求,只會									65

min 'support share difference' on each comments

- ► smaller difference in support share → consensus
- Consensus arguments
 - Main appeal of people (agreed by both groups)

Results: Dispute Comments

comment-id	author-id	a_agree	a_disagree	a_support_share	b_agree	b_disagree	b_support_share	comment-body
				-0.484848				我覺得計程車身一定要塗裝成黃色的,和其他車輛顏色不同。
								我覺得在徹集意見前,各級相關政府單位都應該先明確表示立場·
								我覺得任何的創新服務以違聲利目的的確是社會進步的重要過程,但必須完全在法律的規範下合法經營,
							0.818182	我覺得所有營業車輛都應該一律採取乘客評分機制,而不是只靠政府核發營業許可·
								我覺得 UberX 現行不法行為應盡一切努力使其停業,不需要國人表示意見。

max 'support share difference' on each comments

- larger difference in support share dispute
- Dispute arguments
 - Need to be discussed and solved

Next step

- Dynamic analysis
 - time-series hypothesis testing
 - 4 sub-graphs, calculate the social distance of opinion group A and B

Limitation

- clustering method produce different result
- testing leiden-graph method

Here are some slides that have examples of different syntax you may find helpful (not required to do things like a tikz diagram)

Example of loading a figure you've uploaded (can be pdf or png)



Example of a table

Name	Estimand	Estimator	Why?
Divergence between sample and population	$(\bar{X} - E[\hat{X} \mid T = 1]) - (\bar{X} - E[\hat{X} \mid T = 0])$	Regress distance between population mean (\bar{X}) and sample mean (\hat{X}) on treatment indicator	Measures whether treat- ment produces sample quantities closer to po- pulation mean
Response rate	$\frac{\frac{1}{n}\sum_{\{i:S_i=1\}}Y_i(T)}{1) - Y_i(T) = 0}$	Regress response on treatment	In combination with bi- as measure helps us un- derstand whether we in- crease both response ra- te and decrease non- response bias, or only increase response rate with no reductions in bi- as

Example of inserting code snippet using fragile environment and listings

```
1 def clean_yelp_json(one_biz):
2
     ## restrict to str cols
3
      d_str = \{key: value for key, value in one_biz.items()\}
4
     if type(value) = str }
5
      df_str = pd.DataFrame(d_str, index = [d_str['id']])
6
      return ( df_str )
7
8
9 yelp_stronly = [clean_yelp_json(one_b)
          for one_b in yelp_genjson['businesses']]
yelp_stronly_df = pd.concat(yelp_stronly)
```

Example of splitting slide using minipage and tikz diagram

Random targeting:

Risk-based targeting



based on risk/importance

Another example tikz diagram



Fit top model to data containing training units' 2015, 2017, <u>and 2019</u> features (aggregated so one prediction per unit)	Use that model to predict 2021 nonresponse
per unit)	nonresponse