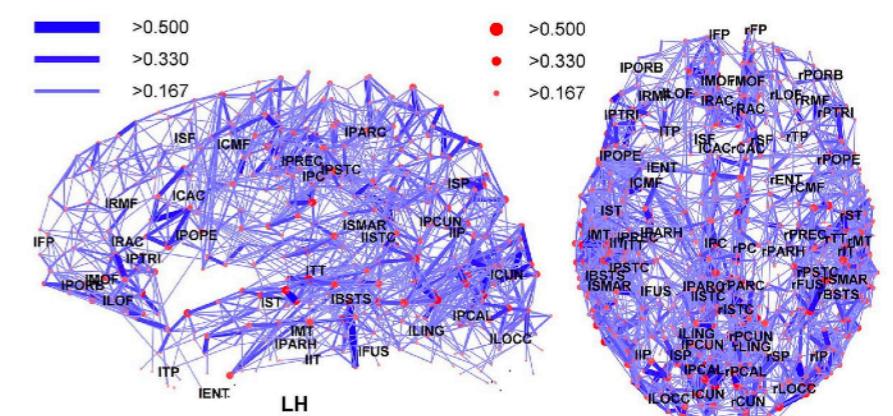
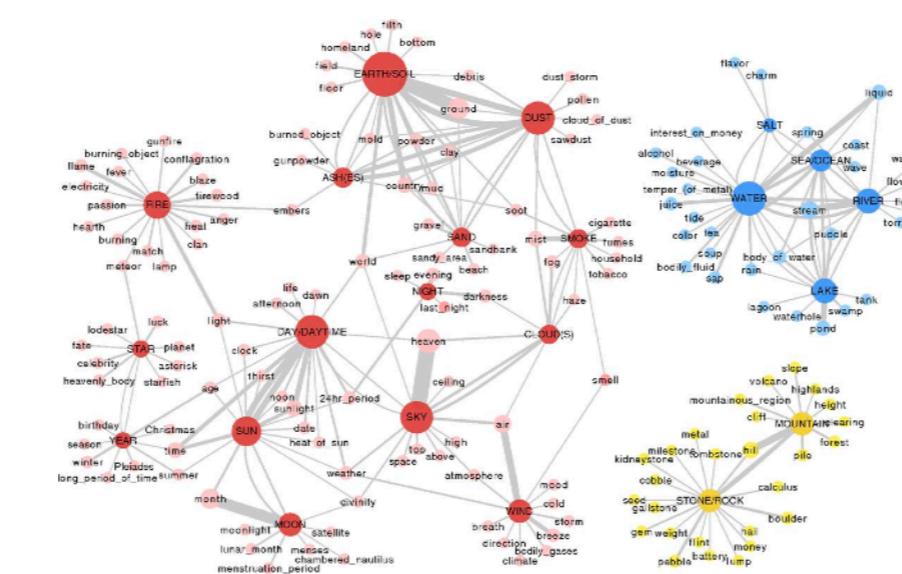
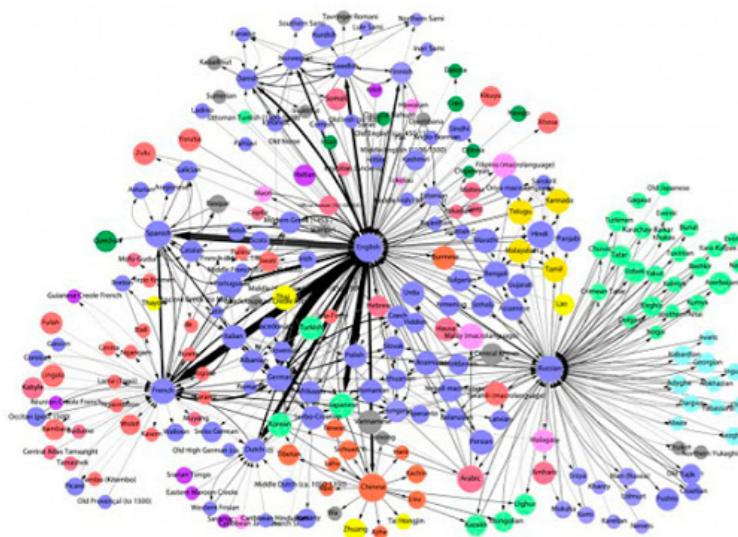
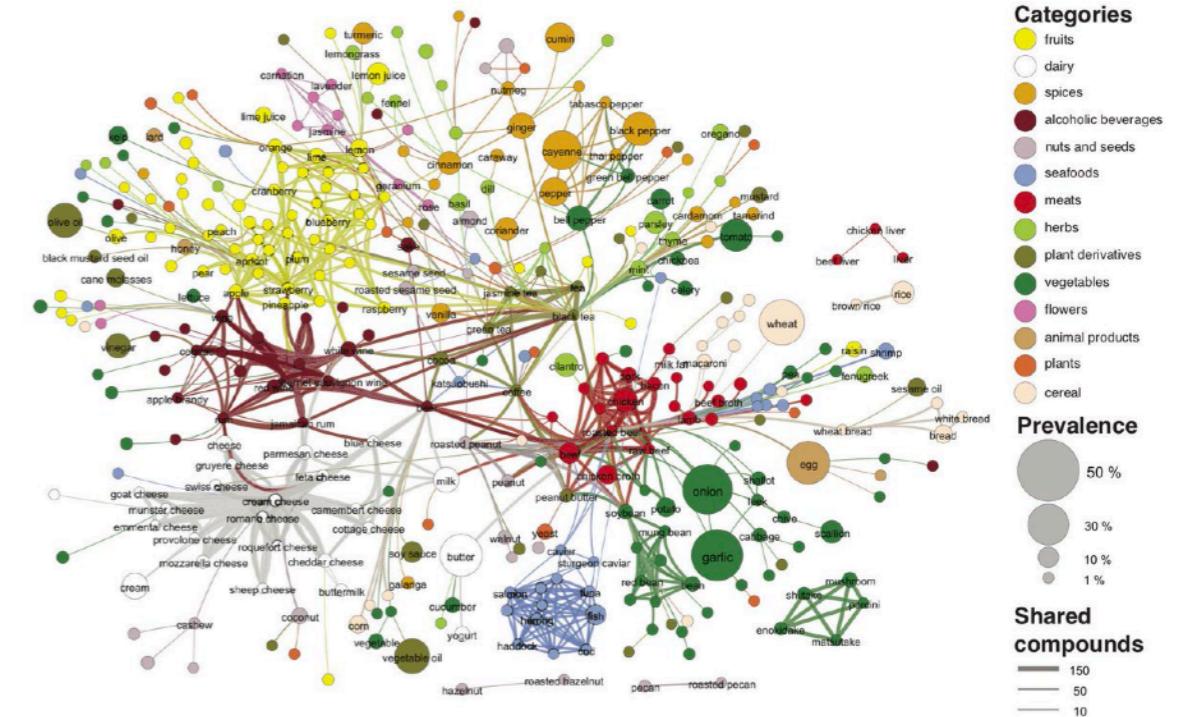
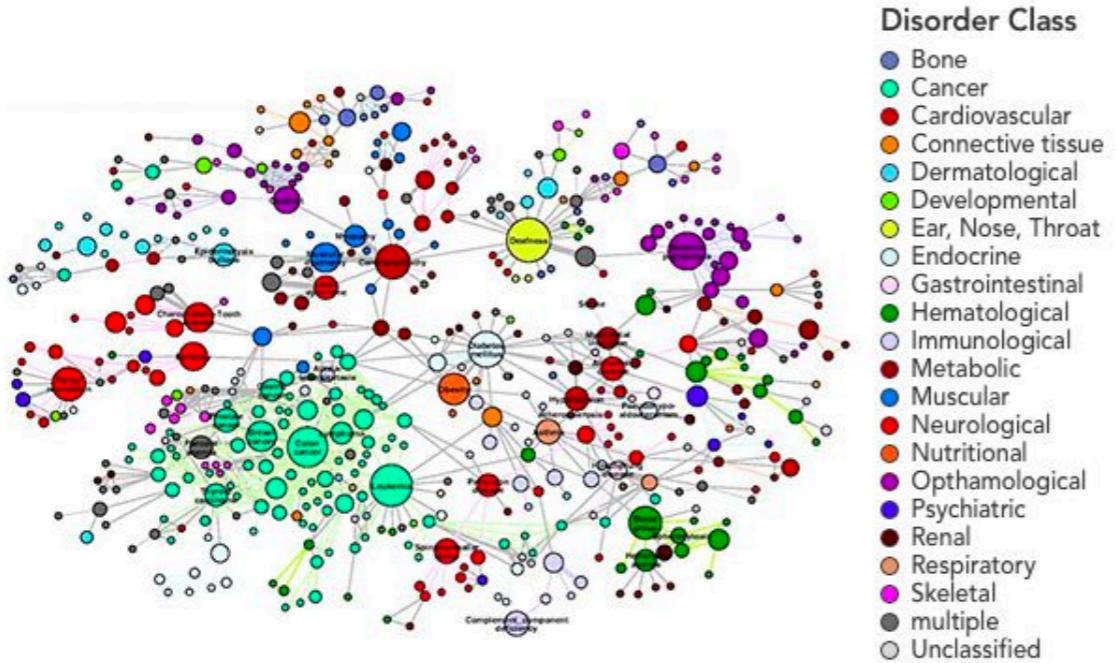
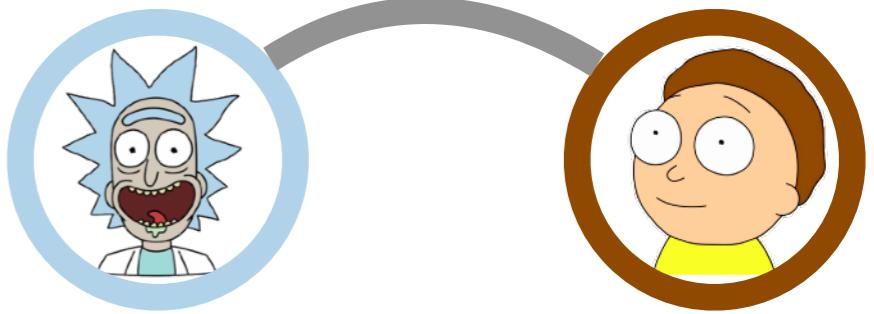


# Identifying Networks of Social Bots on Twitter Conversations

**SICSS - Istanbul'20**

**Onur Varol**  
Sabanci University  
[@onurvarol](https://twitter.com/onurvarol)

# Why networks?



# Online polarization and differences in the communication preferences

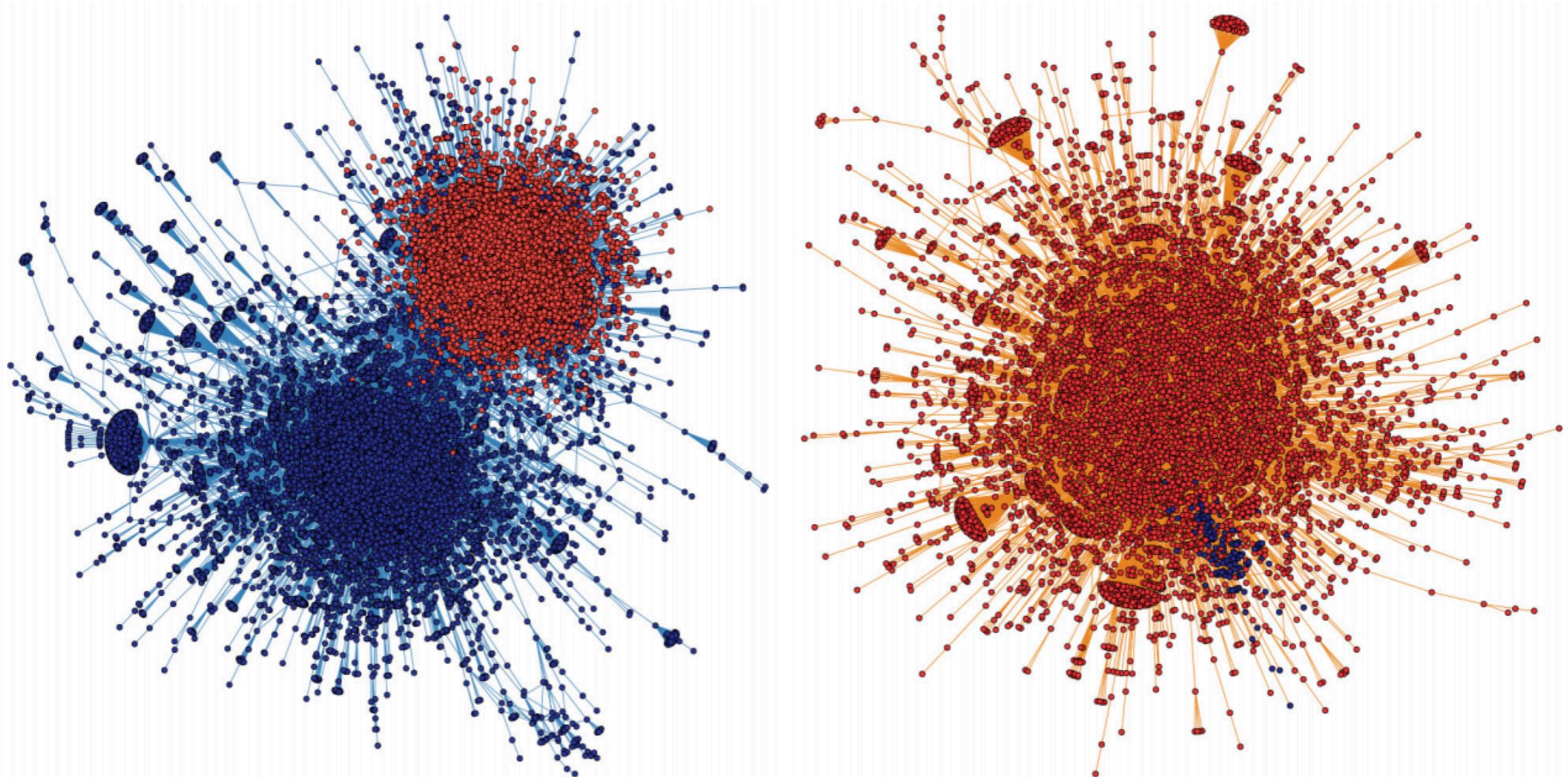


Figure 1: The political retweet (left) and mention (right) networks, laid out using a force-directed algorithm. Node colors reflect cluster assignments (see § 3.1). Community structure is evident in the retweet network, but less so in the mention network. We show in § 3.3 that in the retweet network, the red cluster A is made of 93% right-leaning users, while the blue cluster B is made of 80% left-leaning users.

Public speeches



Abraham Lincoln  
1860

First American  
president on radio



Warren G. Harding  
1922

First political advertisement  
aired on TV



Dwight D. Eisenhower  
1952

1650



1837



1876



1902



1927



Harry S. Truman  
1947



First presidential  
speech on TV

1960

Kennedy & Nixon



First presidential  
debate on TV



## I am Barack Obama, President of the United States -- AMA

POLITICS

submitted 4 years ago \* by PresidentObama

Obama



Hi, I'm Barack Obama, President of the United States. Ask me anything. I'll be taking your questions for half an hour starting at about 4:30 ET.

Proof it's me:

<https://twitter.com/BarackObama/status/240903767350968320>



President Obama

@POTUS



Following

Hello, Twitter! It's Barack. Really! Six years in, they're finally giving me my own account.

RETWEETS

274,794

LIKES

414,318



11:38 AM - 18 May 2015

2012



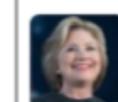
1988



2004



2015



Hillary Clinton

@HillaryClinton



Following

2016



Donald J. Trump

@realDonaldTrump

Obama just endorsed Crooked Hillary. He wants four more years of Obama—but nobody else does!

RETWEETS

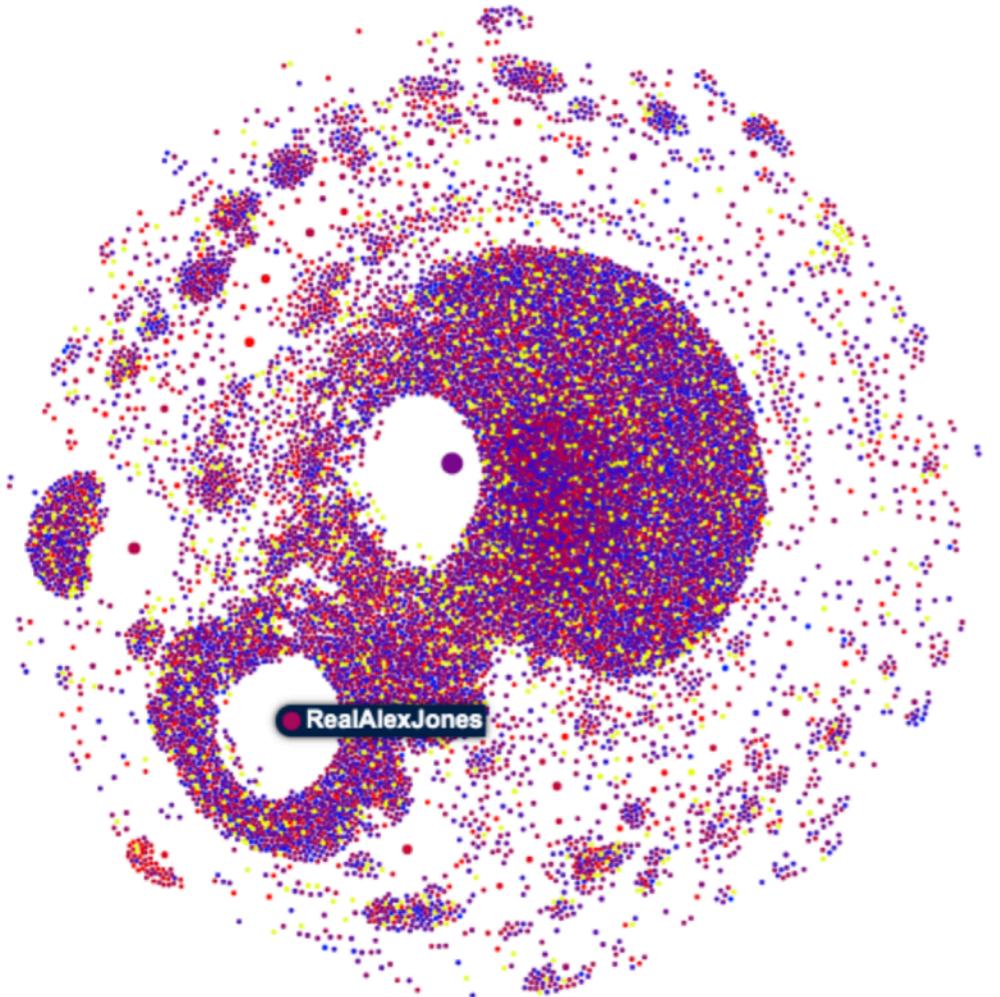
507,017

LIKES

671,876



2:27 PM - 9 Jun 2016



## The spread of a fake news on Twitter

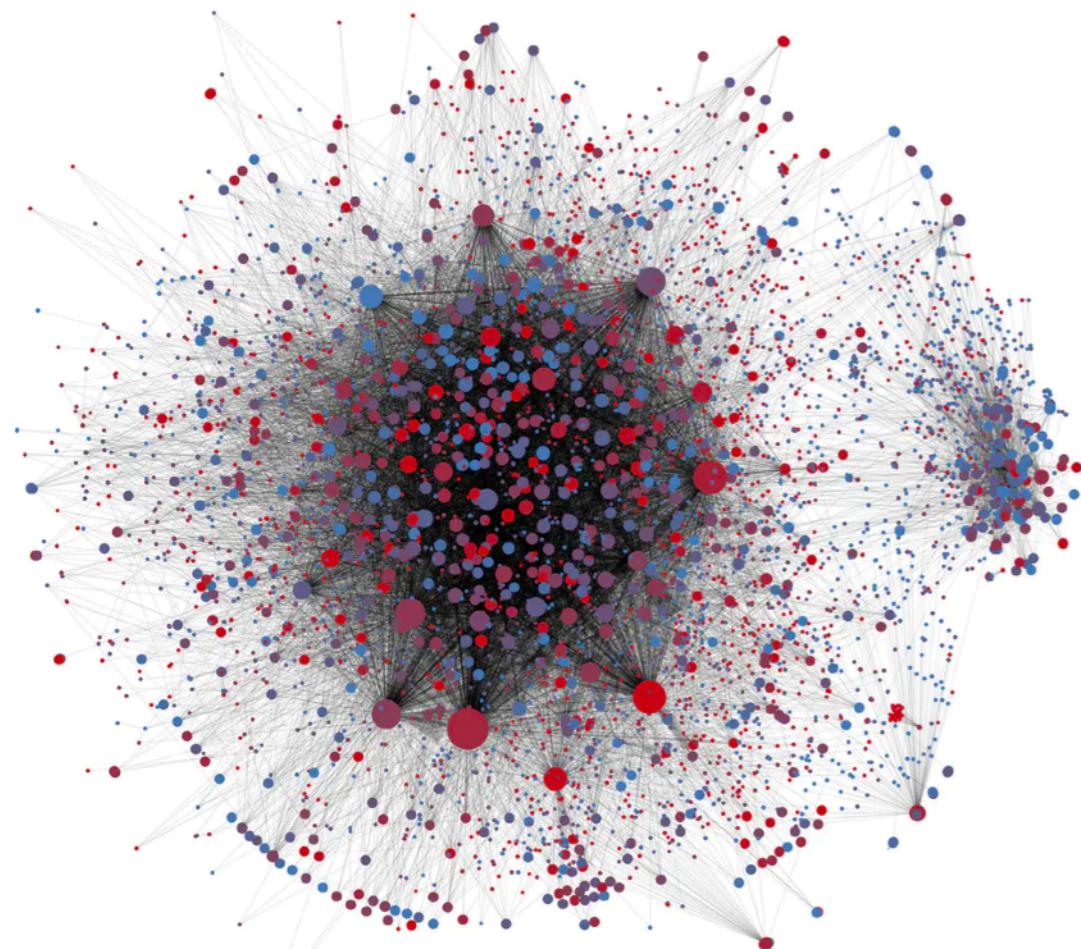
Network visualization of highly viral fabricated news report titled ["Spirit cooking": Clinton campaign chairman practices bizarre occult ritual](#), published 4 days before the 2016 US election and shared in over 30,000 tweets.

Node colors indicate Botometer score and the size of the nodes corresponds to number of retweets account received.

Interactive visualization of this network available

<https://iunetsci.github.io/HoaxyBots/>

Shao, Chengcheng, Giovanni Luca Ciampaglia, Onur Varol, Kai-Cheng Yang, Alessandro Flammini, and Filippo Menczer. [\*\*"The spread of low-credibility content by social bots."\*\*](#) *Nature Communications* 9, no. 1 (2018): 1-9.



## The spread of anti-vaccine messages

How bots are used to affect, and possibly manipulate, the online debate about vaccination policy. It is the retweet network for the #SB277 hashtag, about a recent California law on vaccination requirements and exemptions.

Nodes represent Twitter users, and links show how information spreads among users. The node size represents influence (times a user is retweeted), the color represents bot scores.

Ferrara, Emilio, Onur Varol, Clayton Davis, Filippo Menczer, and Alessandro Flammini. [\*\*"The rise of social bots."\*\*](#) *Communications of the ACM* 59, no. 7 (2016): 96-104.

# BTS Army fans take over racist hashtags

Bots are known for their massive automated activities to flood organic conversations. For the first time an account like @bts\_twt mobilize their 26.5M followers to post content using racist hashtags to make those conversation channels useless.

This is an interesting example of information warfare.

우리는 인종차별에 반대합니다.  
우리는 폭력에 반대합니다.  
나, 당신, 우리 모두는 존중받을 권리가 있습니다. 함께 하겠습니다.  
We stand against racial discrimination.  
We condemn violence.  
You, I and we all have the right to be respected. We will stand together.

#BlackLivesMatter

Translate Tweet

2:01 AM · Jun 4, 2020 · Twitter for Android

998.7K Retweets 2M Likes

I will never bash a k-pop stan ever again. They've taken over the #WhiteLivesMatter and #whiteoutwednesday

Hey, look at us

39 10:55 AM - Jun 3, 2020

See Liz's other Tweets

jess 7 AD2 #BLM

@pardonyoongi

#whiteoutwednesday  
we did it. its under kpop now

TikTok @redpinks

13 7:04 AM - Jun 3, 2020

See jess's 7 AD2 #BLM's other Tweets

☆•Nini•☆

@danielamonsua

#whiteoutwednesday is garbage, but we can improve it with kpop

BLACKPINK

355 7:52 AM - Jun 3, 2020

65 people are talking about this

uriOne

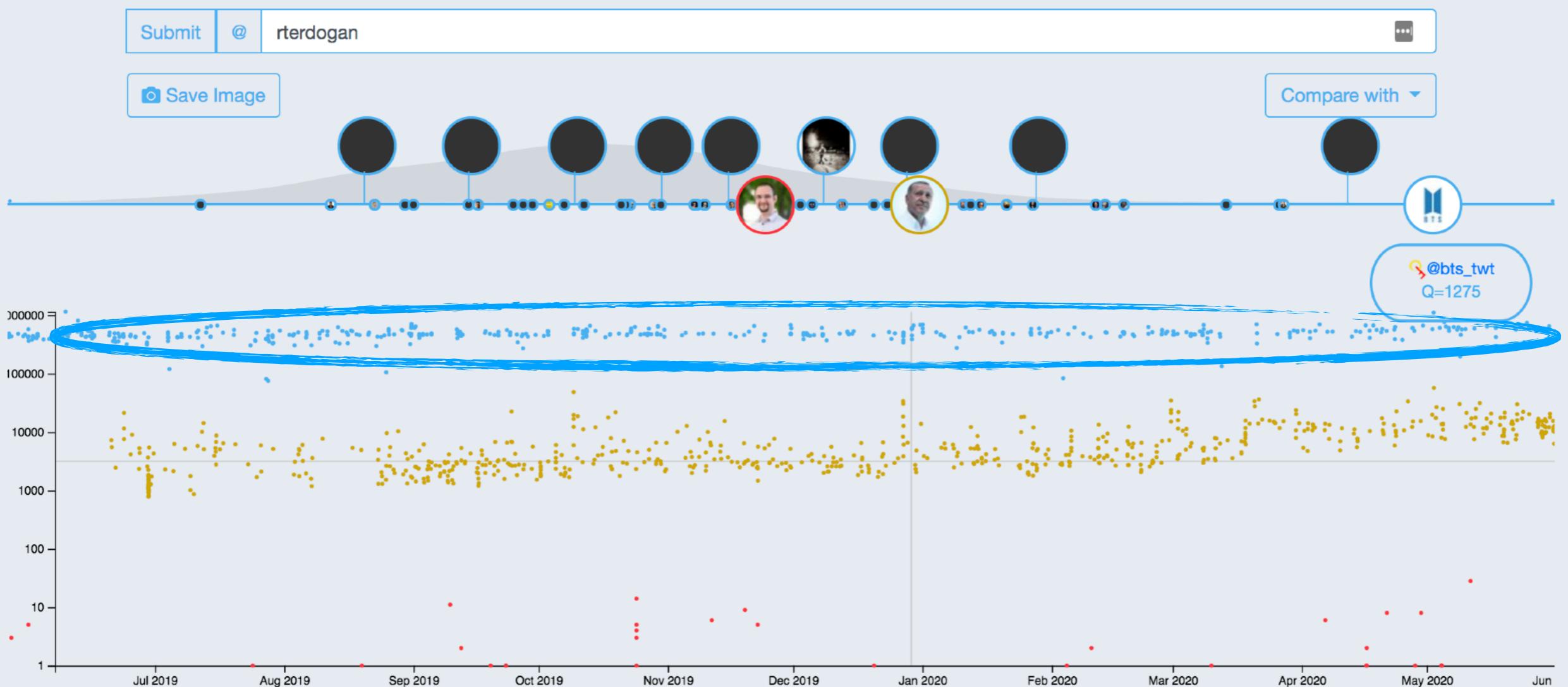
maybe if y'all stanned bts rather than being racists

#WhiteLivesMatter

seokmos

186 1:36 PM - Jun 3, 2020

23 people are talking about this



Anything @bts\_twt send gets nearly 500k retweets.

**Onur Varol**  
@onurvarol

Joining @SabanciU in 2021.  
Previously @NUnetsi & @IndianaUniv -- @MSFTResearch Intern'14 & '15  
Developed @Botometer  
#networkscience, #datascience

① Boston, MA ② onurvarol.com ③ Joined September 2009

3,320 Following 2,610 Followers

**Recep Tayyip Erdogan** ✓  
@RTErdogan

Turkey Cumhurbaşkanı ve AK Parti Genel Başkanı - President of Turkey and AK Party Chairman  
Translate bio

① Ankara, Turkey ③ Joined August 2009

95 Following 16.2M Followers

**BTS** ✓  
@BTS\_twt

Hi! We are BTS!!

② btsblog.ibighit.com ③ Joined July 2011

136 Following 26.5M Followers

# Demo: Select an interesting content to analyze

Let's take a look at a hashtag that we might be interested in analyzing. I selected “**MilliHesaplarBurada**” since there has been recent activity on Twitter.

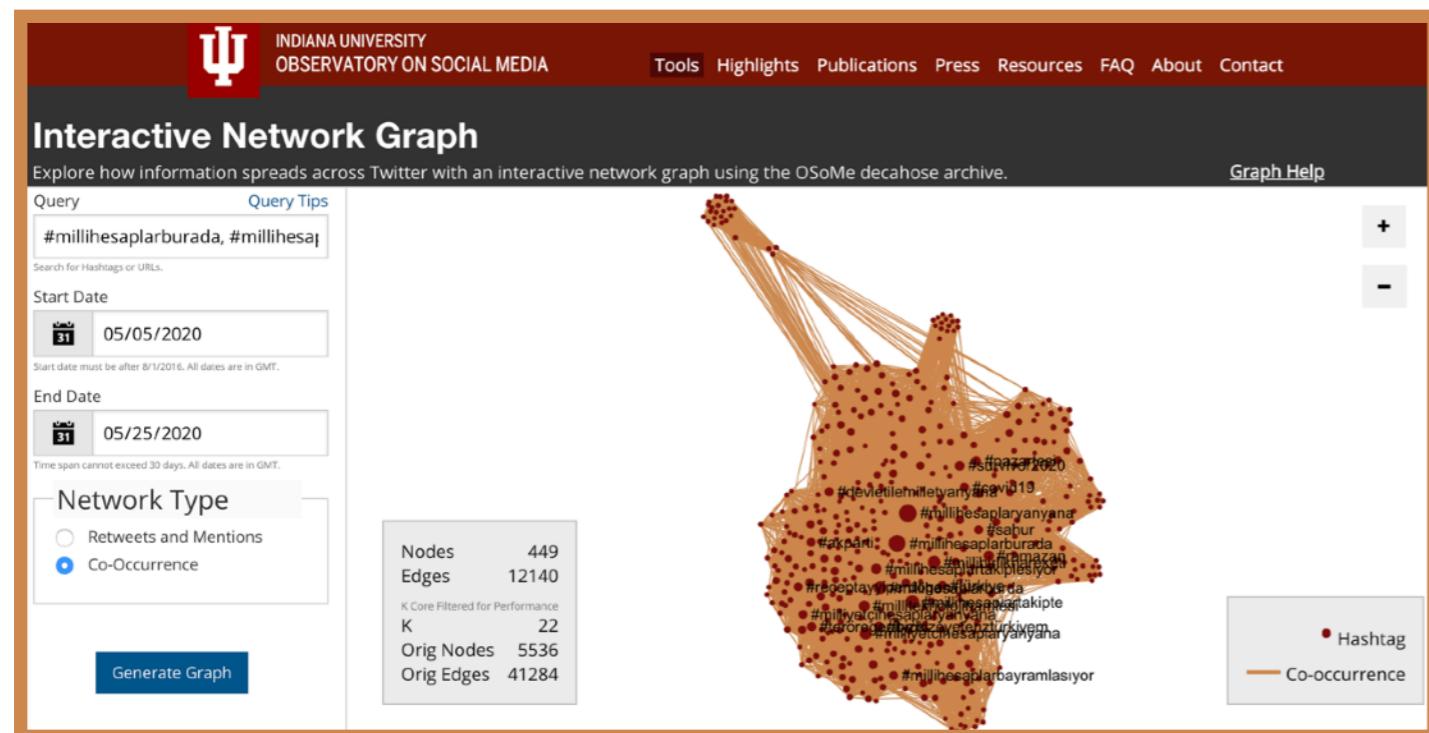
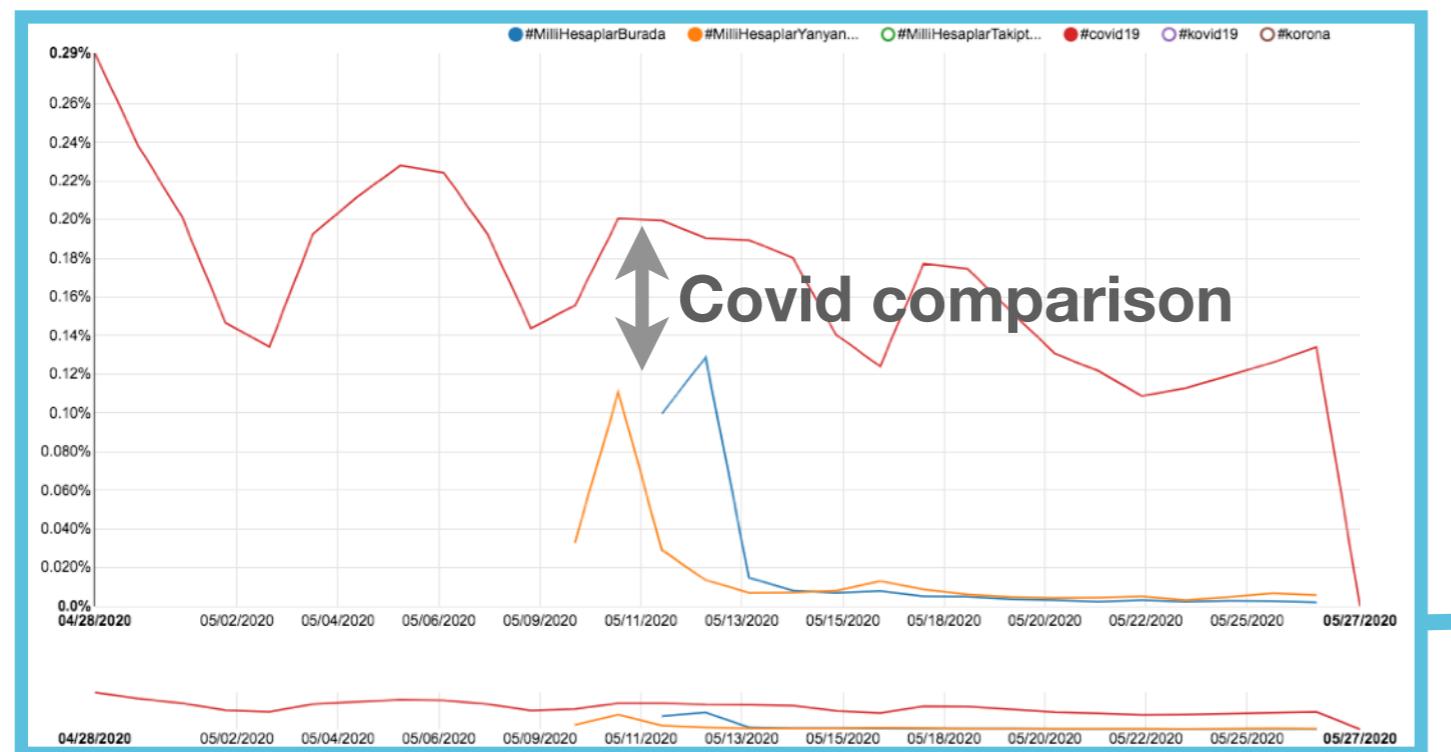
Keçeli @bediuzzamann · May 27  
Milli hesaplarımızın güçlenmesi için sayın Cumhurbaşkanımızın talimatıyla başlayan #MilliHesaplarYanyana çalışmasına destek veriyor, yorum 🇹🇷 bırakıyorum. Sizi takip ediyor, takibinizi bekliyorum.  
#MilliHesaplarBurada

Barbaros Hayrettin @Barbaro59619872 11h  
Takipe takip  
#MilliHesaplarYanyana  
#MilliHesaplarBurada  
#MilliHesaplarTakipte  
#MilliTakipSözü  
#BizBizeYeterizTürkiyem

Barbaros Hayrettin  
@Barbaro59619872  
Joined May 2020  
10 Following 9 Followers

Follow

# Demo: Preliminary data exploration on OSoMe



INDIANA UNIVERSITY  
OBSERVATORY ON SOCIAL MEDIA

Tools Highlights Publications Press Resources FAQ About Contact

Tools (beta)

Misinformation Tools

Tools created by the OSoMe team to help combat the spread of misinformation.

### Hoaxy



Visualize the spread of claims and fact checking.

### Botometer



Check how bot-like a Twitter user behaves.

### Fakey



Play this game to learn to recognize fake news on your social feed.

### BotSlayer



Set up your own system to detect coordination and bot amplification on Twitter.

### EchoDemo



Simulation demonstrating how two basic mechanisms of social media can lead to polarized social networks.

### Bot Electioneering Volume



Visualize the activity of likely bots on Twitter around the 2018 US midterm elections.

OSoMe Tools

Tools that use the OSoMe Decahose data archive.

### Trends



Compare when memes gain and lose popularity.

### Networks



Explore who is discussing a meme and what memes are related.

### Maps



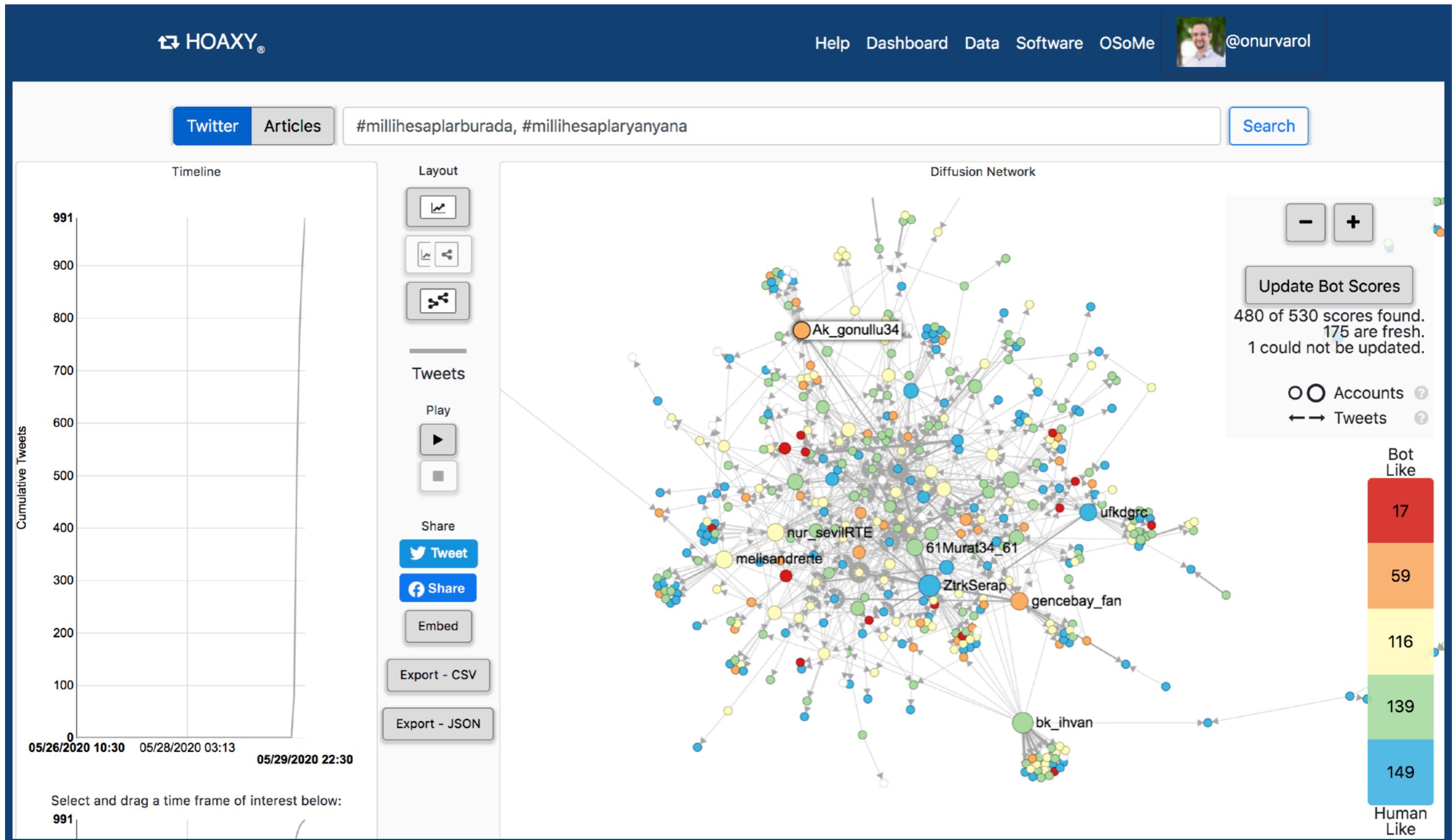
Examine where people are talking about a meme over time.

### Movies



Generate movies of how conversations about a meme evolve over time.

# Demo: Look at diffusion patterns on Hoaxy



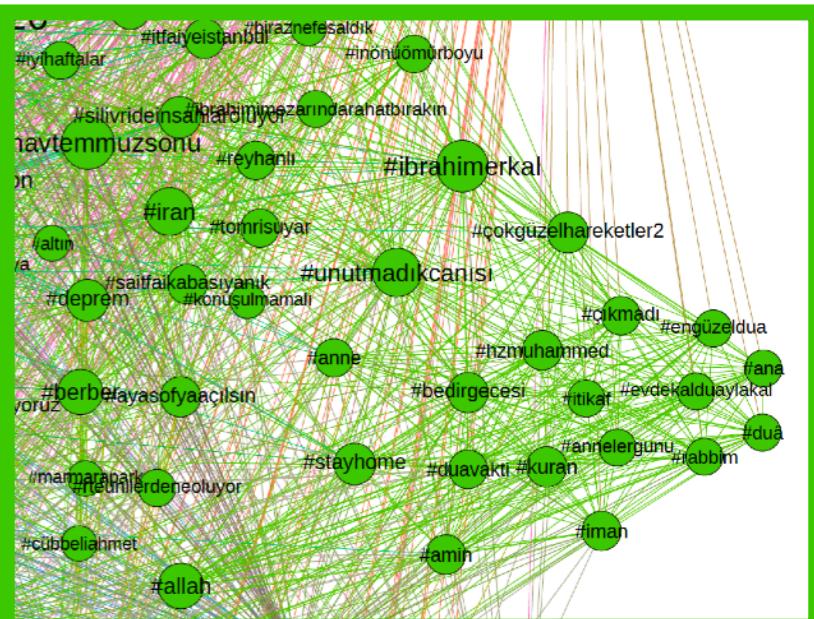
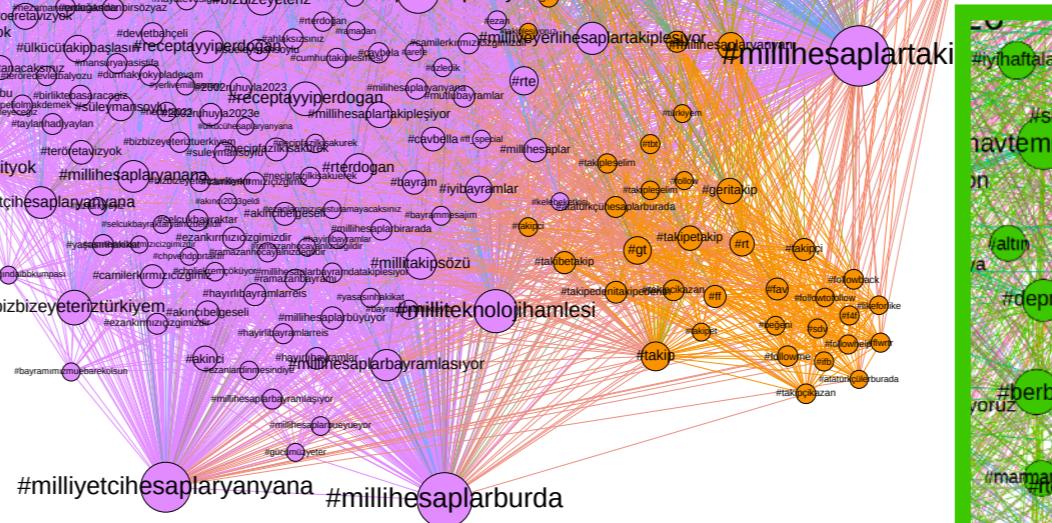
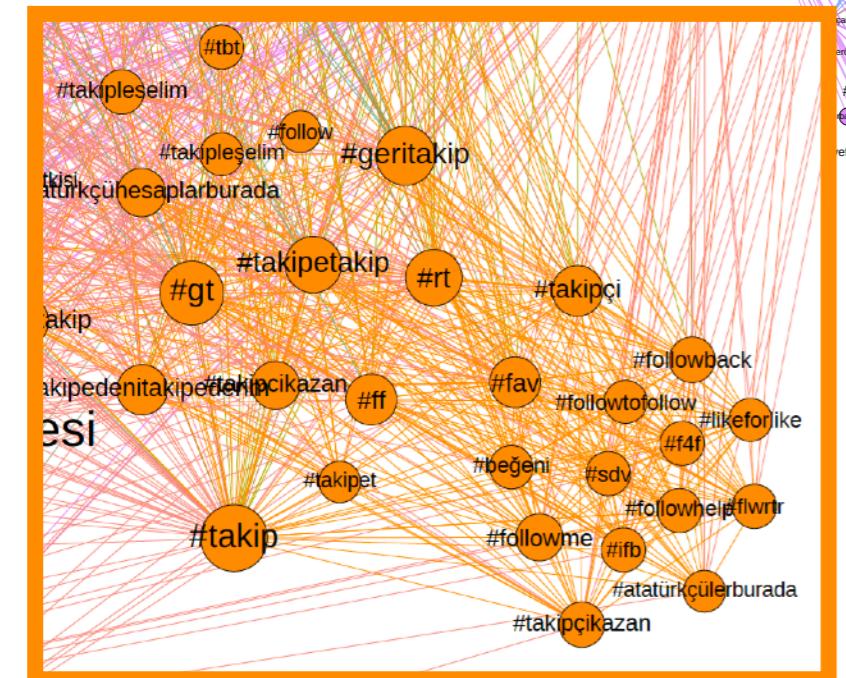
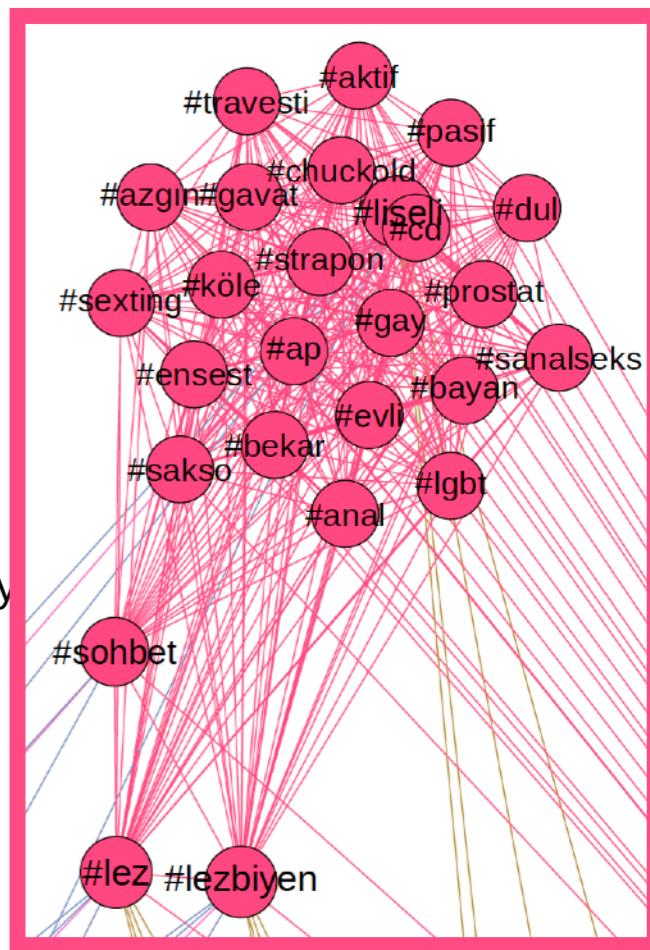
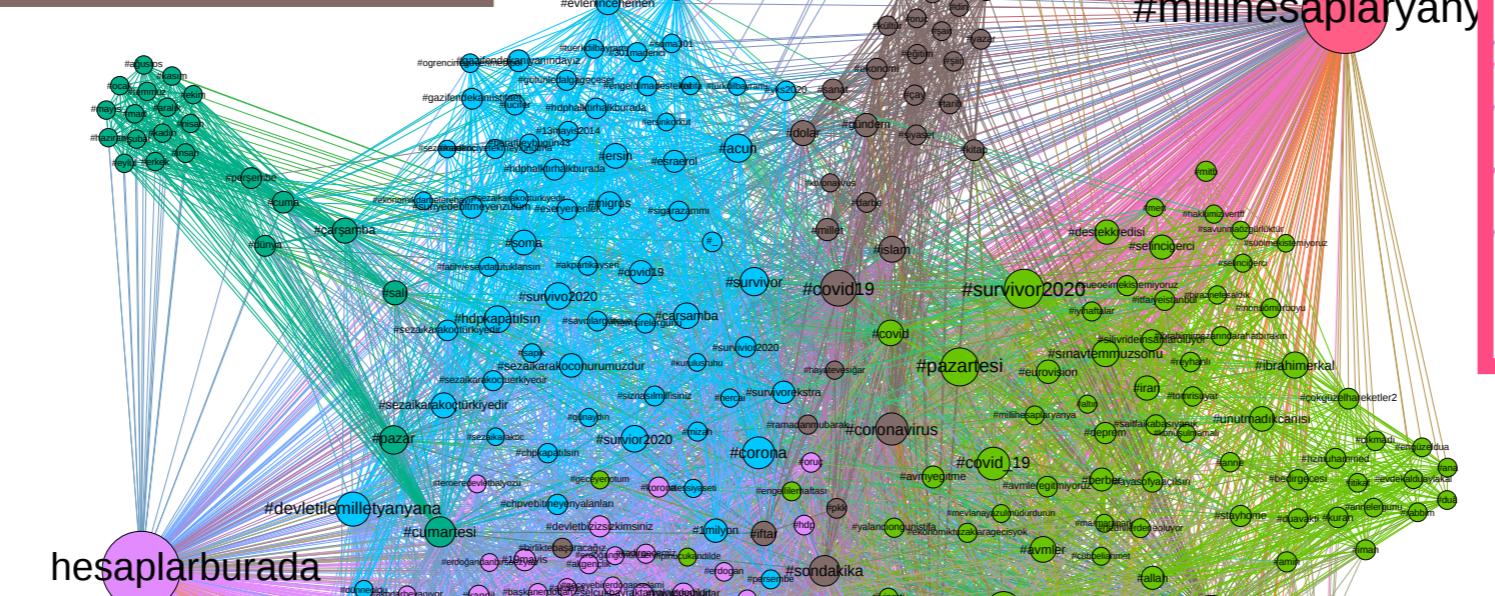
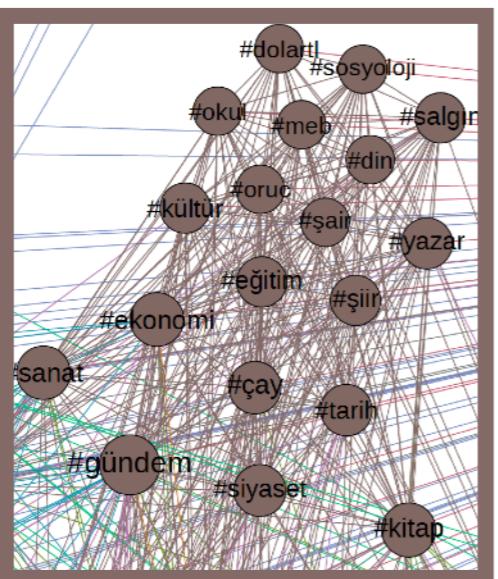
Shao, Chengcheng, et al. "Hoaxy: A platform for tracking online misinformation." *Proc. of the 25th Intl. Conf. Companion on World Wide Web*. 2016.

# **Demo: Preliminary data analysis using OSoMe export**

Let's take a look at

**1\_analyze-hashtag-cooccurrence.ipynb**

notebook for the analysis



	degree_N	degree_C	closeness_C	betweenness_C	pagerank_N	clustering_N	clique_N	core_N	onion_N
#millihesaplaryanya	428	0.955357	0.957265	0.270459	0.020445	0.095395	27	27	57
#millihesaplarburada	382	0.852679	0.871595	0.157603	0.017969	0.106100	24	27	57
#millihesaplartakipte	276	0.616071	0.722581	0.046616	0.012718	0.146561	24	27	57
#millihesaplarburda	244	0.544643	0.687117	0.032099	0.011220	0.164069	19	27	57
#milliyetcihesaplaryanya	213	0.475446	0.655930	0.022223	0.009810	0.184782	22	27	57
#millihesaplartakiplesiyor	164	0.366071	0.612022	0.011403	0.007525	0.221832	19	27	57
#milliveyerlihesaplartakiplesiyor	106	0.236607	0.567089	0.004340	0.004931	0.288230	19	27	56
#millihesaplaryana	103	0.229911	0.564943	0.003467	0.004791	0.302494	19	27	56
#milliyetcihesaplaryanya	103	0.229911	0.564943	0.003758	0.004846	0.309918	19	27	56
#millihesaplarbayramlaşıyor	101	0.225446	0.563522	0.002991	0.004703	0.327327	19	27	56
#millihesaplaryanyan	74	0.165179	0.545012	0.002133	0.003562	0.370603	24	27	52
#millihesaplartakipleşiyor	67	0.149554	0.540410	0.001100	0.003161	0.425599	19	27	56
#millihesaplar	53	0.118304	0.531435	0.000635	0.002581	0.476778	16	27	55
#millihesaplarbüyüyor	47	0.104911	0.527680	0.000359	0.002325	0.541166	17	27	53
#millihesaplarbirarada	38	0.084821	0.522145	0.000223	0.001915	0.559033	16	27	51
#millihesaplarbayramlaşıyor	32	0.071429	0.518519	0.000101	0.001679	0.697581	16	26	42
#millihesaplarbayramdatakiplesiyor	30	0.066964	0.517321	0.000073	0.001565	0.708046	17	26	41
#millihesaplaryanya	26	0.058036	0.514943	0.000078	0.001422	0.686154	14	23	24
#millihesaplarbueyueyor	25	0.055804	0.514351	0.000035	0.001353	0.790000	15	24	25
#yerlimillihesaplaryanya	24	0.053571	0.513761	0.000103	0.001351	0.579710	11	21	12
#millihesaplaryanyanabiz	15	0.033482	0.508513	0.000026	0.000959	0.761905	12	15	1

**Demo:** Collect content by dehydrating from Tweet IDs

Let's take a look at

**2\_dehydrate-tweetids.ipynb**

notebook for the analysis

# Botometer®

An OSoMe project (bot•o•meter)



Botometer (formerly BotOrNot) checks the activity of a Twitter account and gives it a score based on how likely the account is to be a bot. Higher scores are more bot-like.

Note the following:

- Use of this service requires authenticating with Twitter. For more, see [this FAQ](#).
- Botometer often categorizes "organizational accounts", like [@BarackObama](#), as bot accounts.
- If something's not working, [please let us know](#).

Botometer is a joint project of the Indiana University Network Science Institute ([IUNI](#)) and the Center for Complex Networks and Systems Research ([CNetS](#)).

Interested in using Botometer in your application? Check out our [API](#).

onurvarol Check user Check followers Check friends

Please help us improve Botometer by participating a short, anonymous survey. [Go to the survey ➔](#) X

▼  @onurvarol X

0.5 / 5 

English-specific features	Language-independent features	Bot score based on	<a href="#">?</a>
Content: 1.1		All features: 0.5	
Sentiment: 1.0		Language-independent: 0.4	
		User profile language: en	
		Complete Automation Probability: 0%	

Profile Tweet Details Feedback

English-specific features  
Content: 1.1  
Sentiment: 1.0

Language-independent features  
Friend: 0.7  
Network: 0.5  
Temporal: 0.4  
User: 1.5

Bot score based on  
All features: 0.5  
Language-independent: 0.4  
User profile language: en  
Complete Automation Probability: 0%

Export data

# Feature extraction and datasets

Botometer system trained on using extensive dataset of bot behaviors.

We extract over 1,150 features capturing six subclasses:

- User
- Network
- Friends
- Temporal
- Content
- Sentiment

Features designed to model account behaviors more generic than human-bot difference and can be applied for different classifications problems.

Table 1: List of 1150 features extracted by our framework.

User meta-data	Screen name length	(***)	Happiness scores of aggregated tweets
	Number of digits in screen name	(***)	Valence scores of aggregated tweets
	User name length	(***)	Arousal scores of aggregated tweets
	Time offset (sec.)	(***)	Dominance scores of single tweets
	Default profile (binary)	(*)	Happiness score of single tweets
	Default picture (binary)	(*)	Valence score of single tweets
	Account age (days)	(*)	Arousal score of single tweets
	Number of unique profile descriptions	(*)	Dominance score of single tweets
	(*) Profile description lengths	(*)	Polarization score of single tweets
	(*) Number of friends distribution	(*)	Entropy of polarization scores of single tweets
	(*) Number of followers distribution	(*)	(*) Positive emoticons entropy of single tweets
	(*) Number of favorites distribution	(*)	(*) Negative emoticons entropy of single tweets
	Number of friends (signal-noise ratio and rel. change)	(*)	(*) Emoticons entropy of single tweets
	Number of followers (signal-noise ratio and rel. change)	(*)	(*) Positive and negative score ratio of single tweets
	Number of favorites (signal-noise ratio and rel. change)	(*)	(*) Number of positive emoticons in single tweets
	Number of tweets (per hour and total)	(*)	(*) Number of negative emoticons in single tweets
	Number of retweets (per hour and total)	(*)	(*) Total number of emoticons in single tweets
	Number of mentions (per hour and total)		Ratio of tweets that contain emoticons
	Number of replies (per hour and total)		
	Number of retweeted (per hour and total)		
Friends ( $\dagger$ )	Number of distinct languages		Number of nodes
	Entropy of language use		Number of edges (also for reciprocal)
	(*) Account age distribution	(*)	Strength distribution
	(*) Time offset distribution	(*)	In-strength distribution
	(*) Number of friends distribution	(*)	Out-strength distribution
	(*) Number of followers distribution		Network density (also for reciprocal)
	(*) Number of tweets distribution		(*) Clustering coeff. (also for reciprocal)
Content	(*) Description length distribution		
	Fraction of users with default profile and default picture		
	(*, **) Frequency of POS tags in a tweet	(*)	Time between two consecutive tweets
	(*, **) Proportion of POS tags in a tweet	(*)	Time between two consecutive retweets
	(*) Number of words in a tweet	(*)	Time between two consecutive mentions
Timing	(*) Entropy of words in a tweet		

<sup>†</sup> We consider four types of connected users: retweeting, mentioning, retweeted, and mentioned.

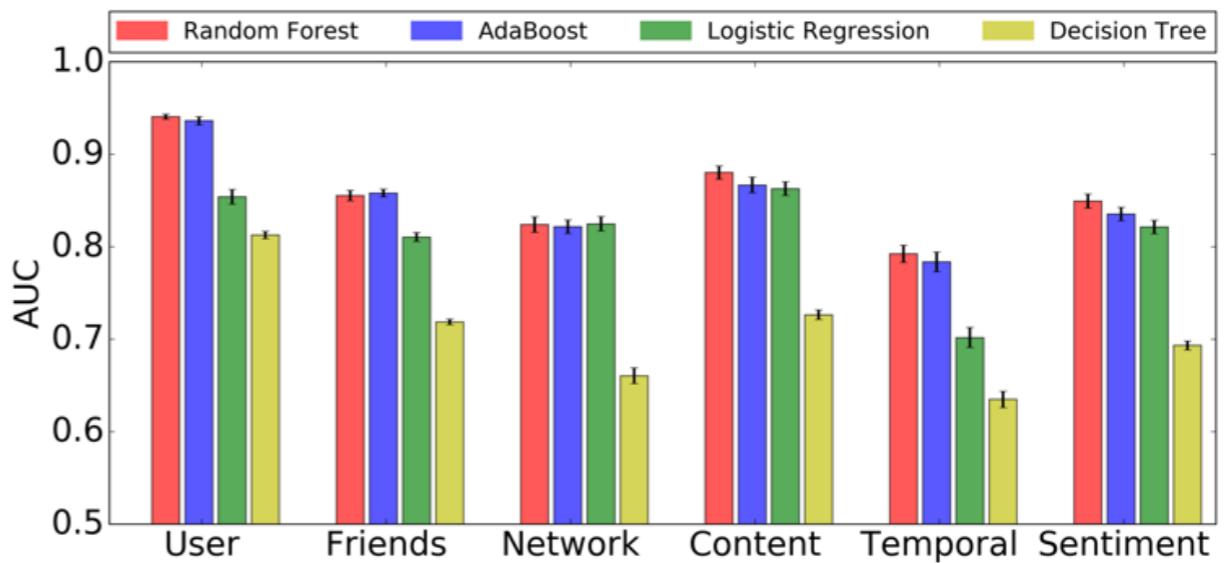
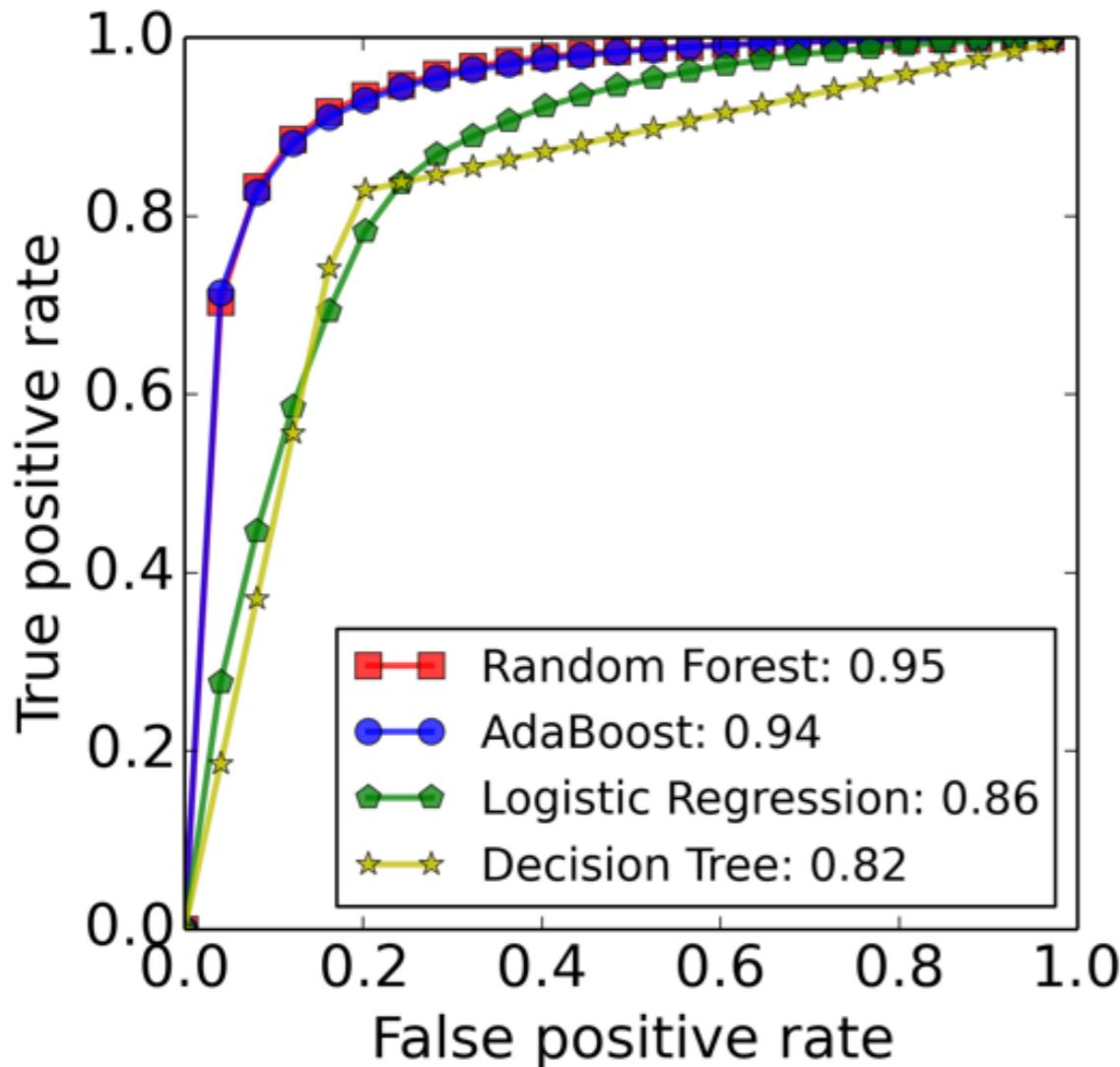
<sup>‡</sup> We consider three types of network: retweet, mention, and hashtag co-occurrence networks.

<sup>\*</sup> Distribution types. For each distribution, the following eight statistics are computed and used as individual features: min, max, median, mean, std. deviation, skewness, kurtosis, and entropy.

<sup>\*\*</sup> Part-Of-Speech (POS) tag. There are nine POS tags: verbs, nouns, adjectives, modal auxiliaries, pre-determiners, interjections, adverbs, wh-, and pronouns.

<sup>\*\*\*</sup> For each feature, we compute mean and std. deviation of the weighted average across words in the lexicon.

# Learning algorithms



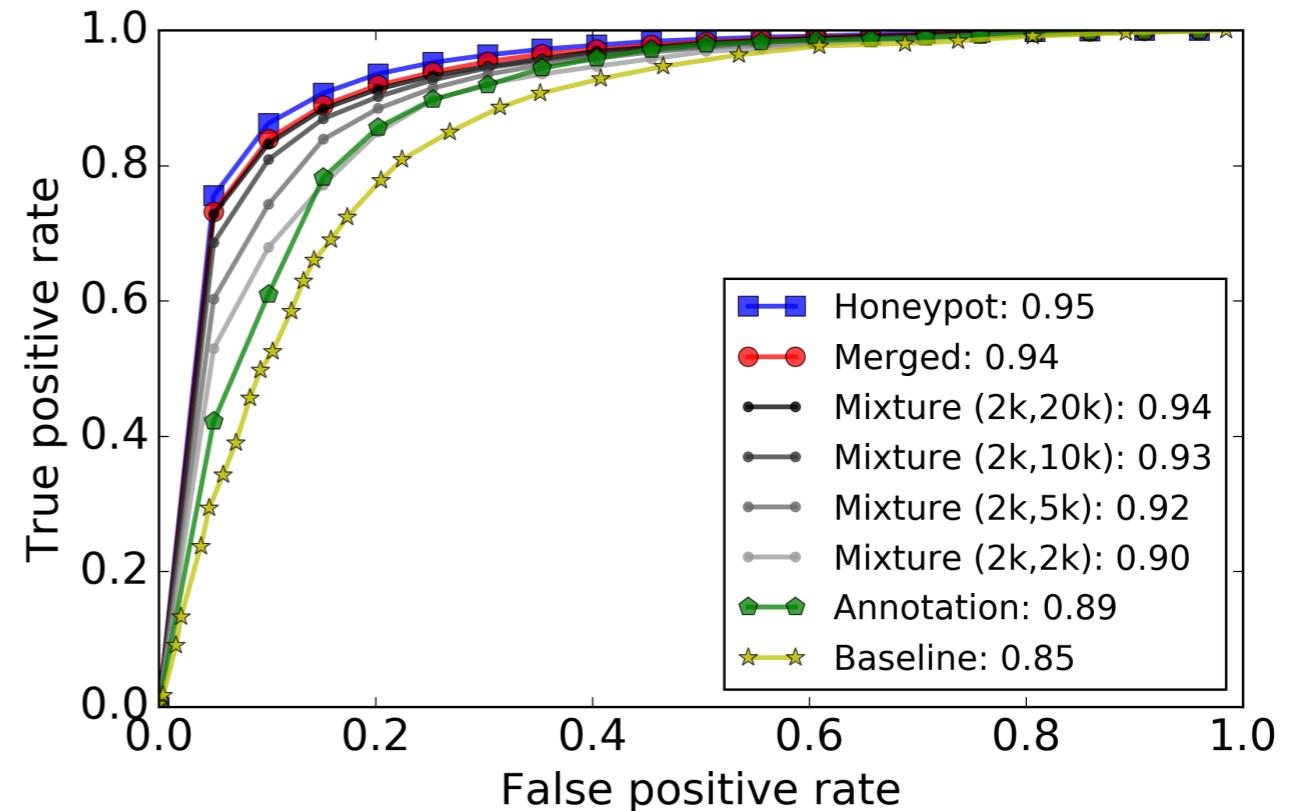
Initial models trained on an existing honeypot dataset<sup>1</sup>

We evaluated contribution of different feature classes and learning models.

[1] Lee, Kyumin, Brian David Eoff, and James Caverlee. "Seven Months with the Devils: A Long- Term Study of Content Polluters on Twitter." ICWSM 2011.

# Estimating bot population by tuning model sensitivity

Trained on mixtures of honeypot and manually annotated data for tuning sensitivity to recent bot behaviors.

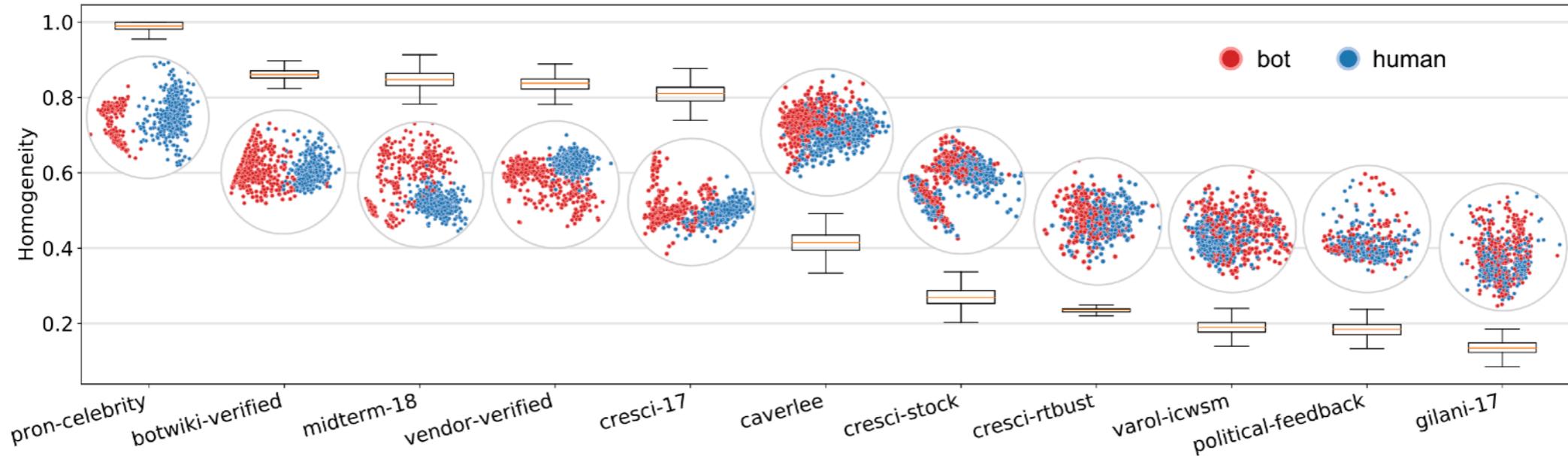


Adjusting sensitivity of detection system for sophisticated bots

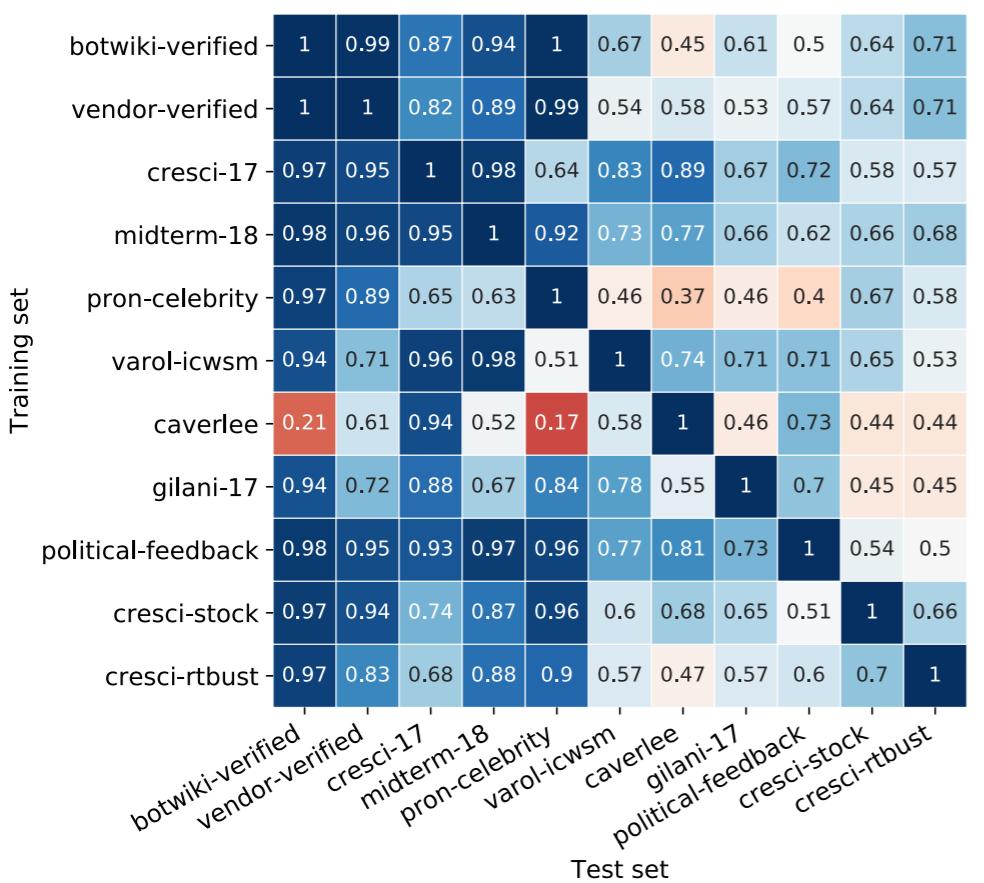


9-15% of active accounts have automated behavior

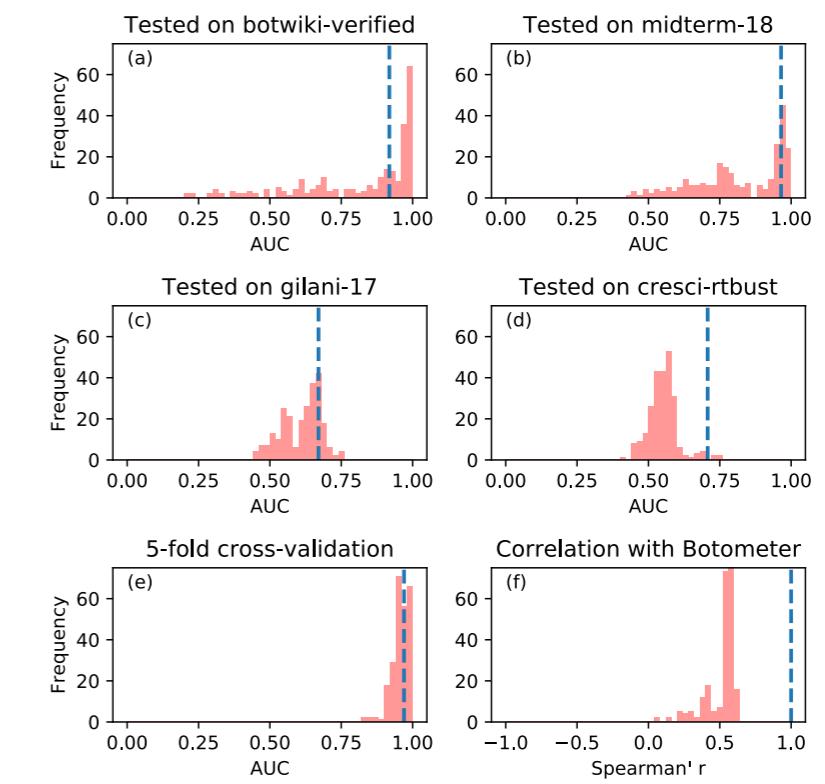
# Improvement via data augmentation



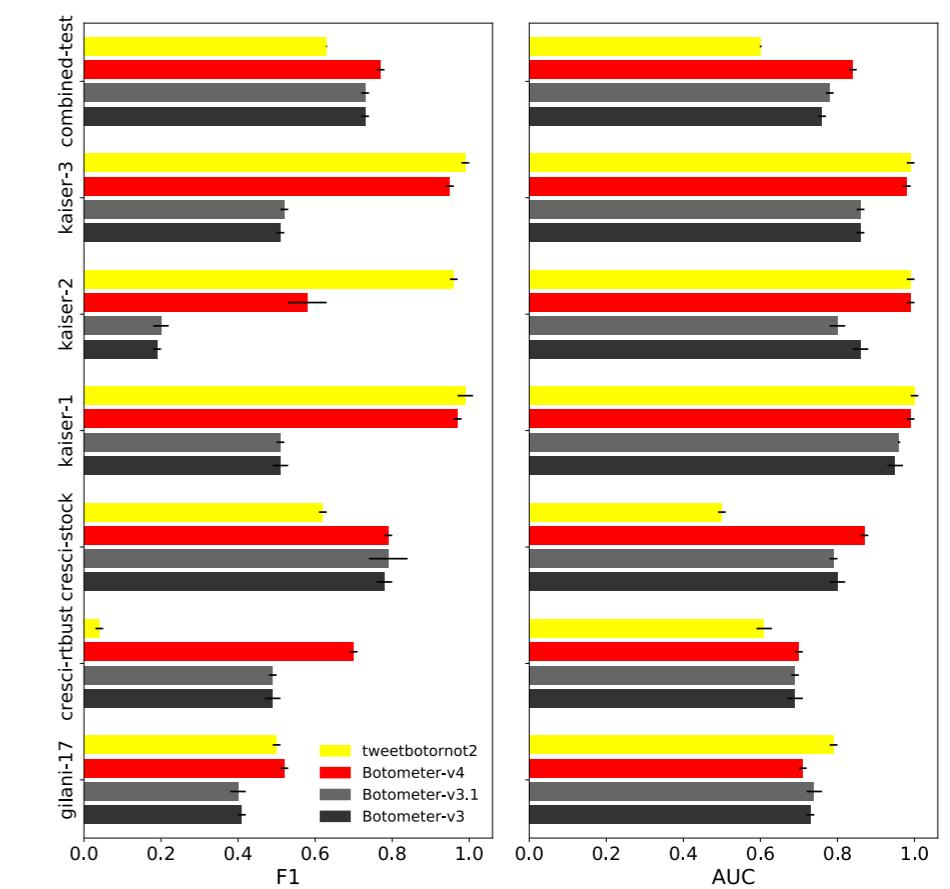
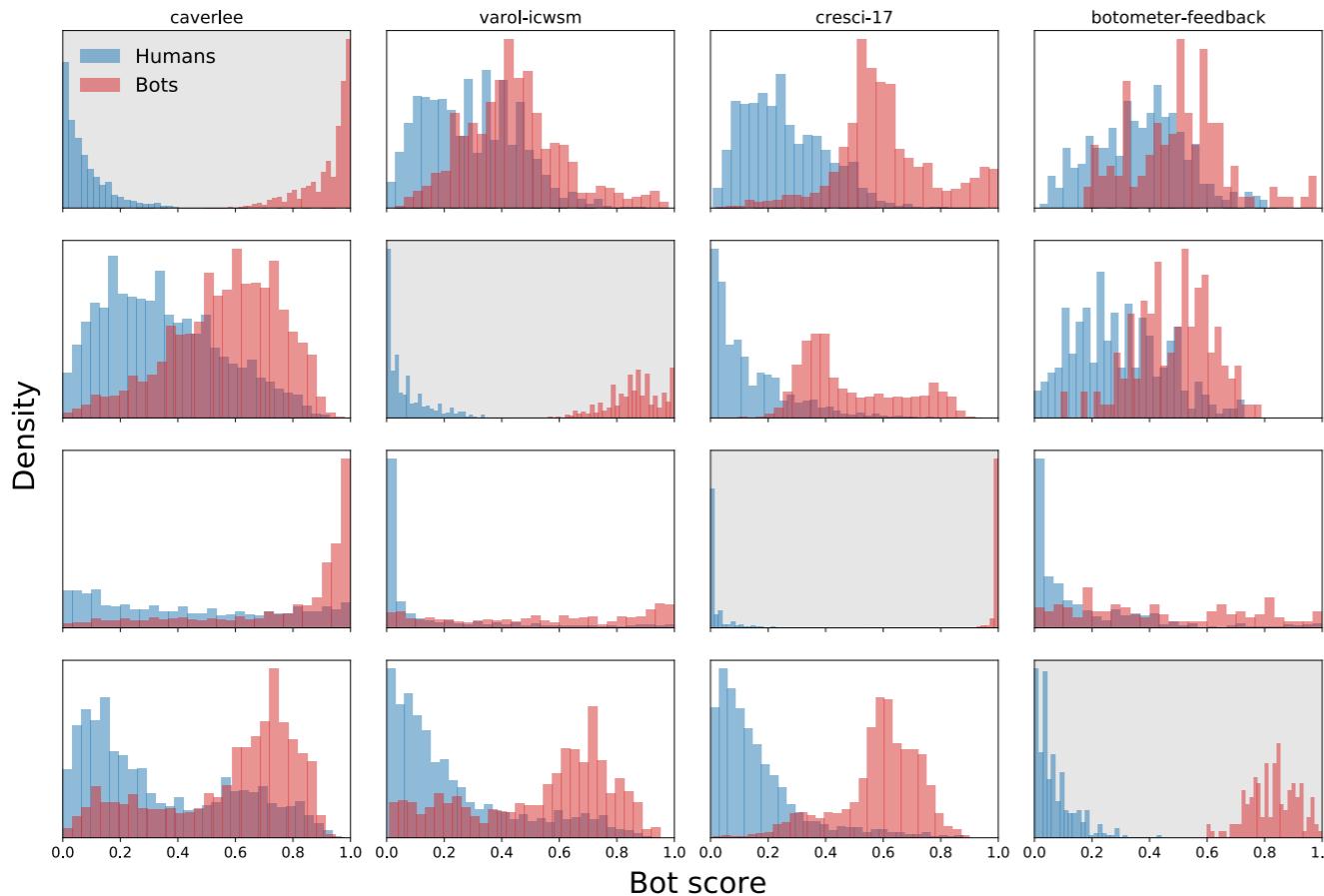
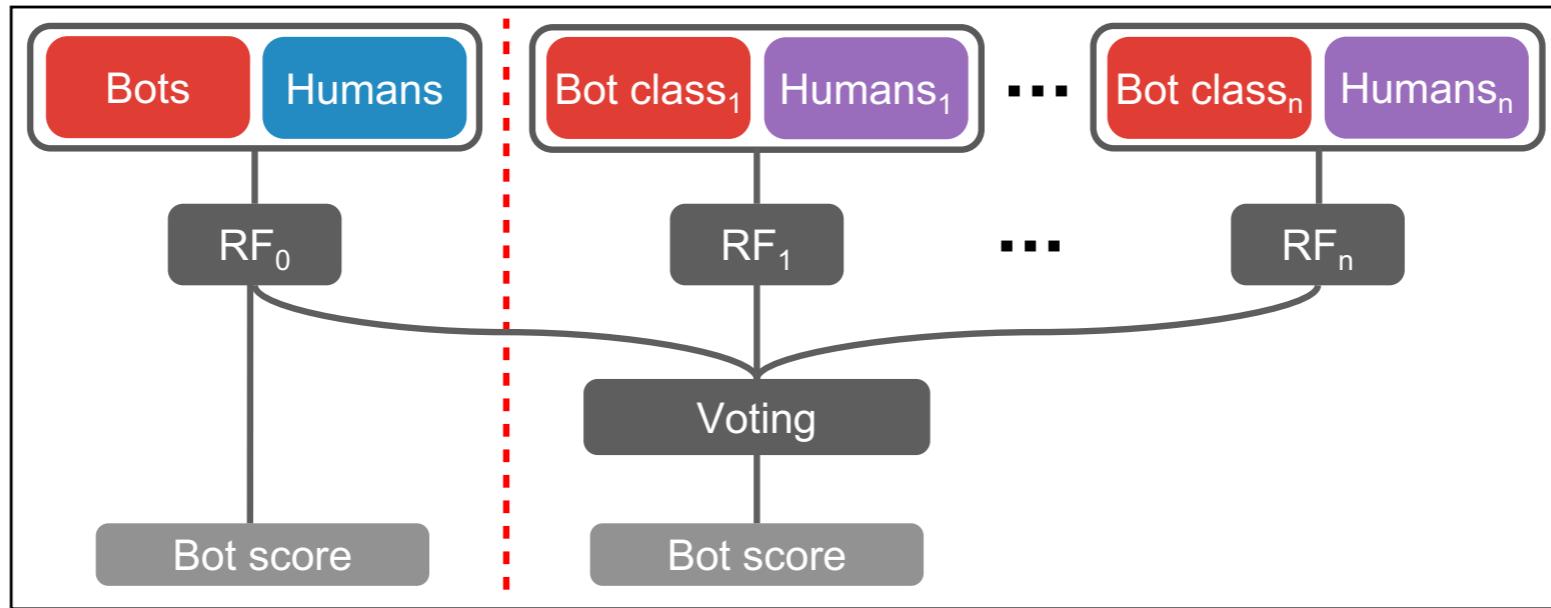
## Out of domain performance



## Model selection



# Improvement via specialized models

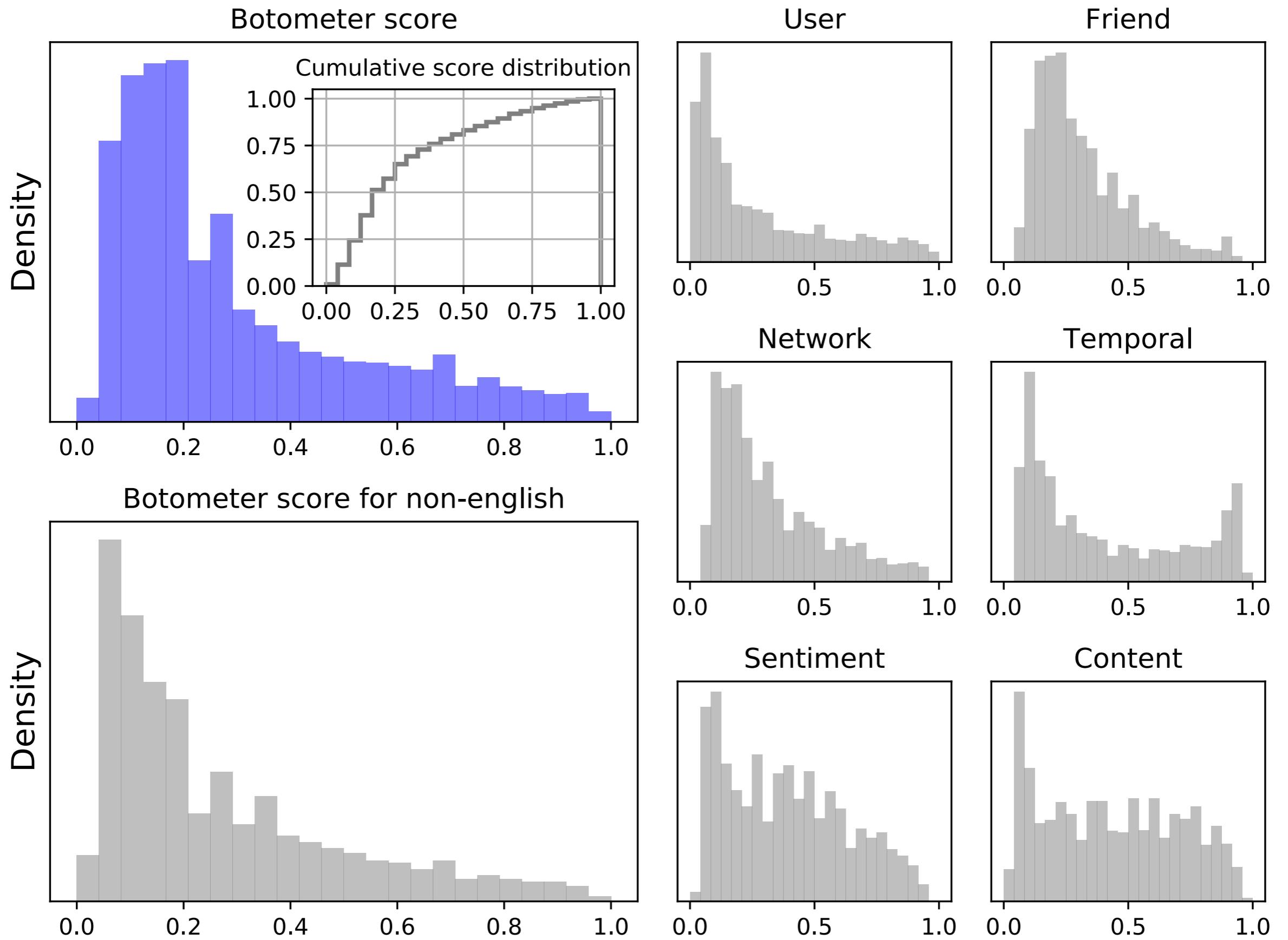


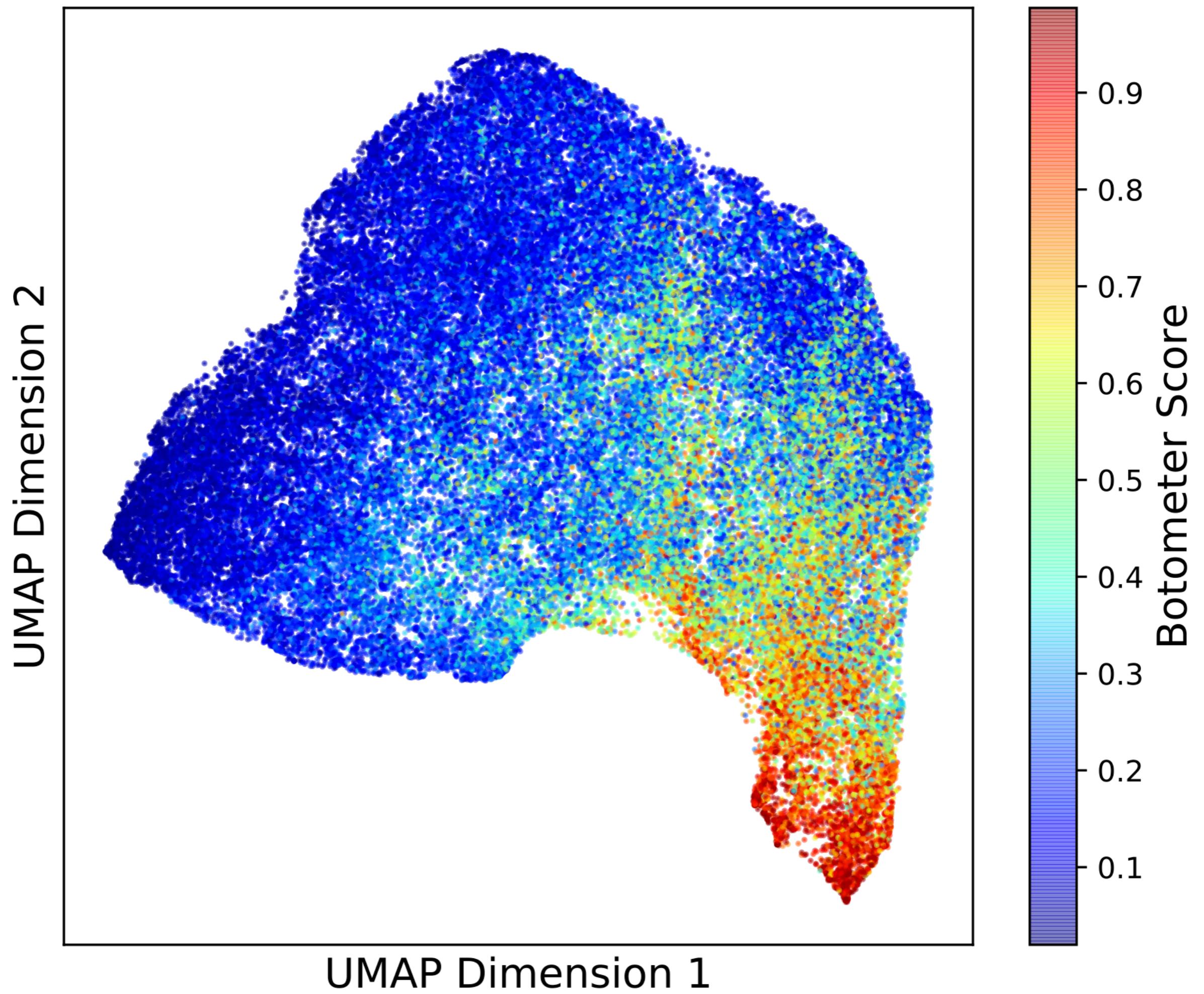
# **Demo: Collecting bot scores using Botometer API**

Let's take a look at

**3\_collect-botometer-scores.ipynb**

notebook for the analysis



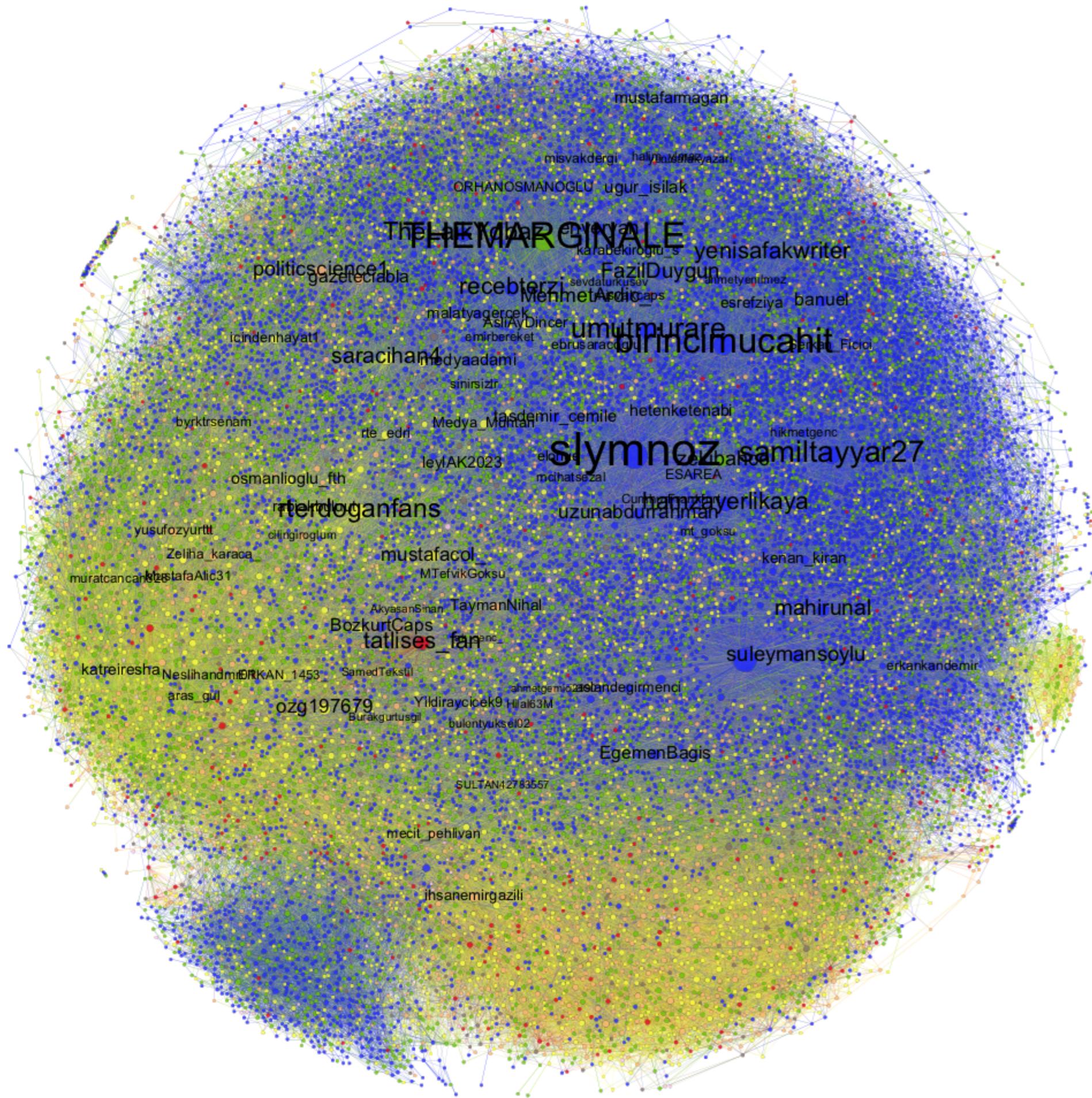


## **Demo: Analyze Twitter networks and Botometer scores**

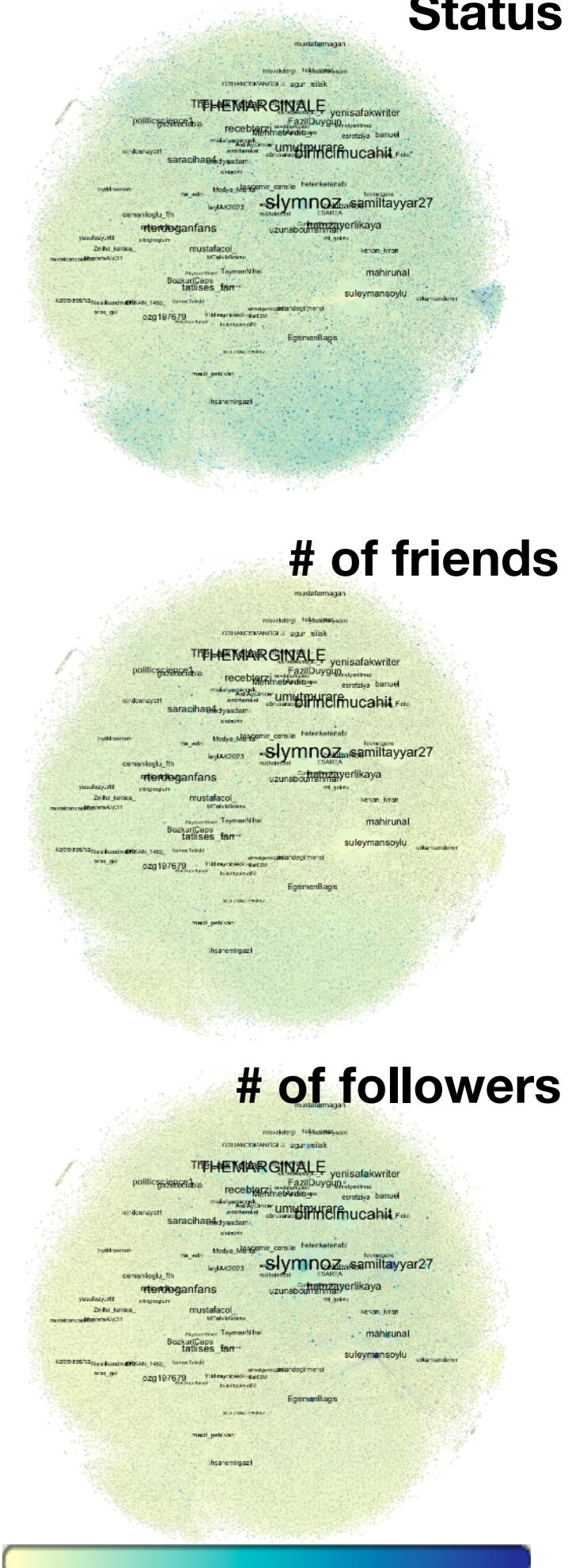
Let's take a look at

**4\_analyze-twitter-networks.ipynb**

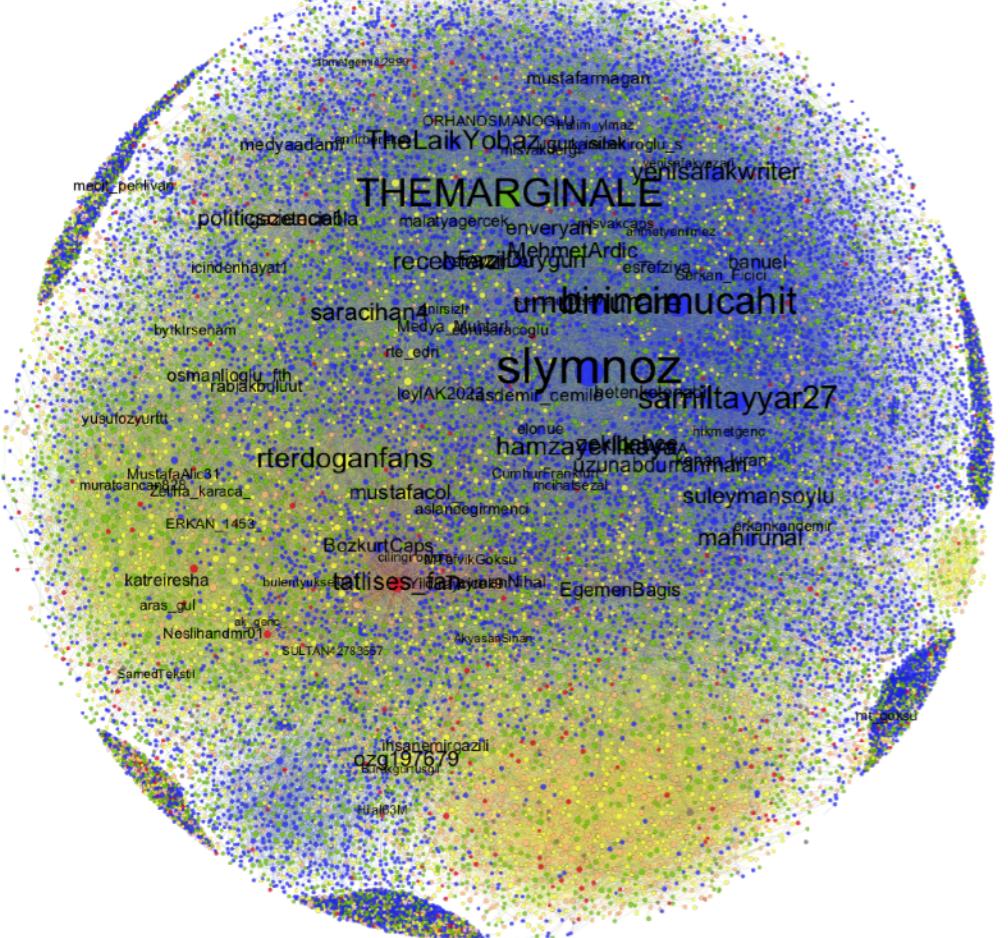
notebook for the analysis



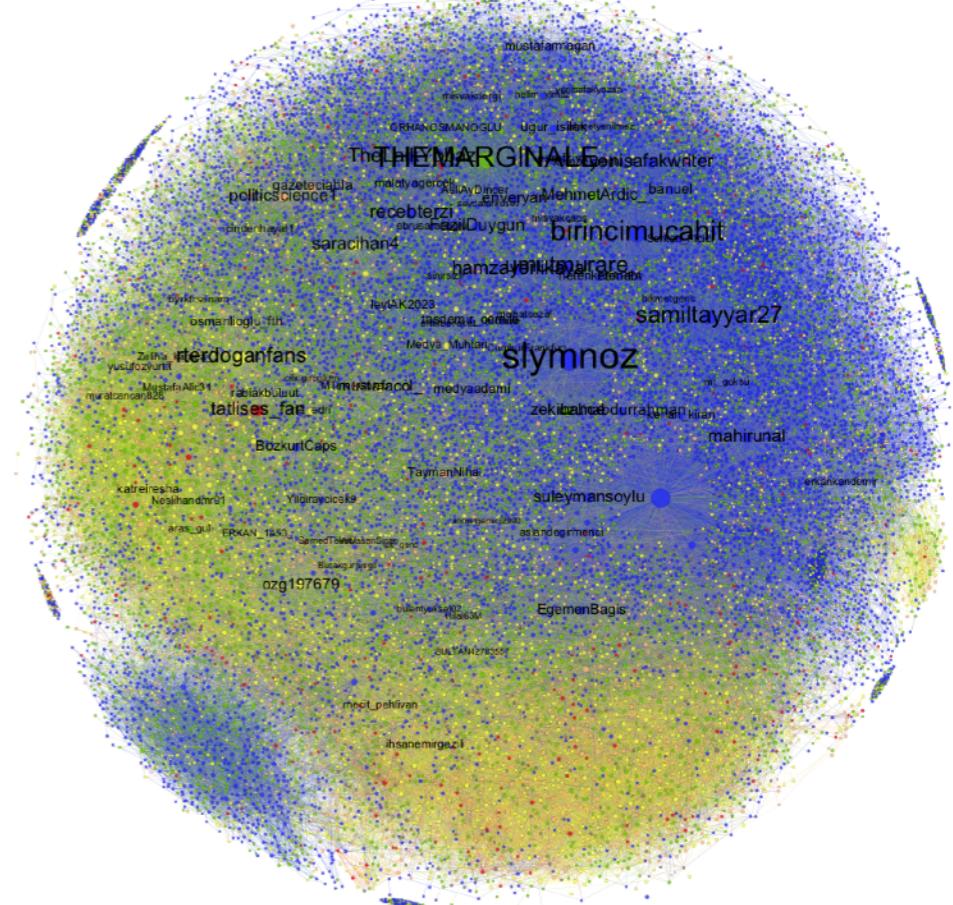
## Status



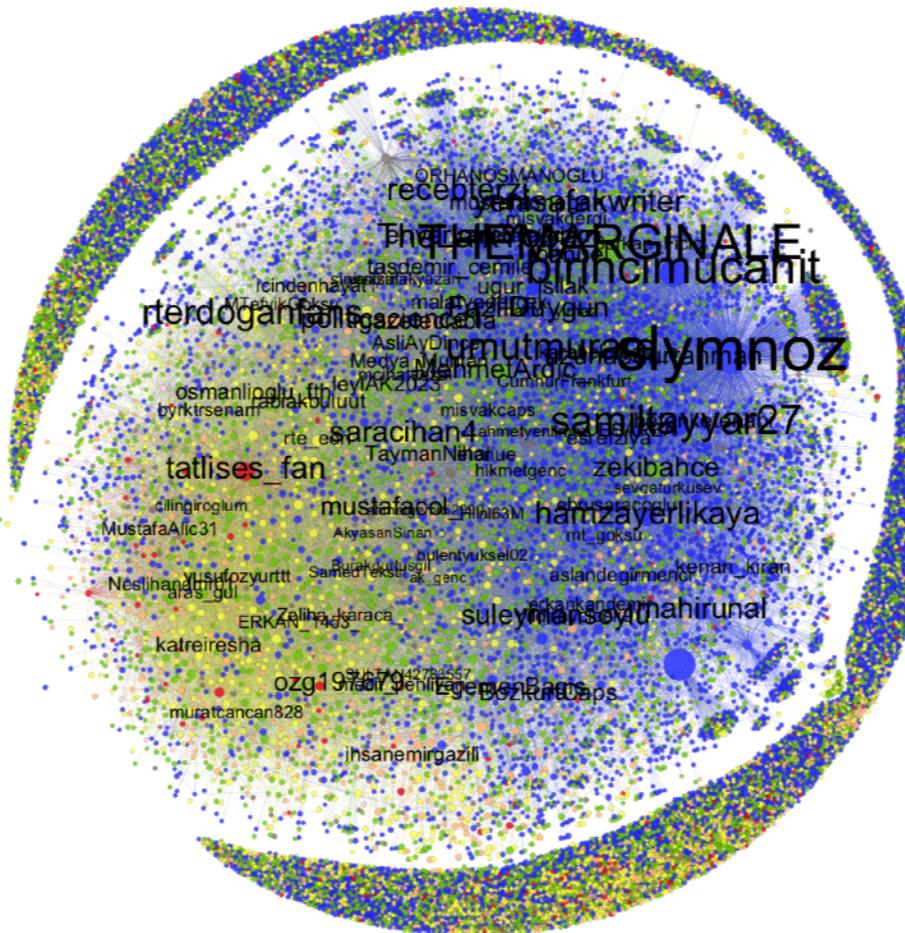
# Retweet



## Mention



## Reply



#millihesaplartakipte

#pazartesi

@sakirbilmez34

@kartalkarapinar

@oguzhanmuezzino  
#akincibelgeseli

#milliteknolojihamlesi  
#hayirlibayramlar  
#reis  
#millihesaplarbayramlaşıyor

@haberulakci

@ucankucukimam

#teroregeçityok

#milliyetçihesaplaryanya

#milliyetçihesaplaryanya

#ülküütakipbaşlasın

@hattapoglu\_

@aasumanerdogan

@behappy5

#türkiv  
#teroregeçityok

#millihesaplarbayramlaşıyor

@avdnmermerci

@karadenizli0808

@in\_turken

@akalin2eyup

#millihesaplaryanya

#millihesaplarburada

#millihesaplarburada

#millihesaplartakiplesiyor

#millihesaplaryanya

#millihesaplaryanya

#bizbizeyeteriz

#millihesaplarbüyüyor

#millitakipsözü

@erkanmut7

#bizbizeyeterizturkiyem

#millibirlikhareketi

#millihesaplartakiplesiyor

#millihesaplartakiplesiyor

#receptayyiperdogan

#ff

#akinci2023geliyor  
#ekonomikdarbelerehayir

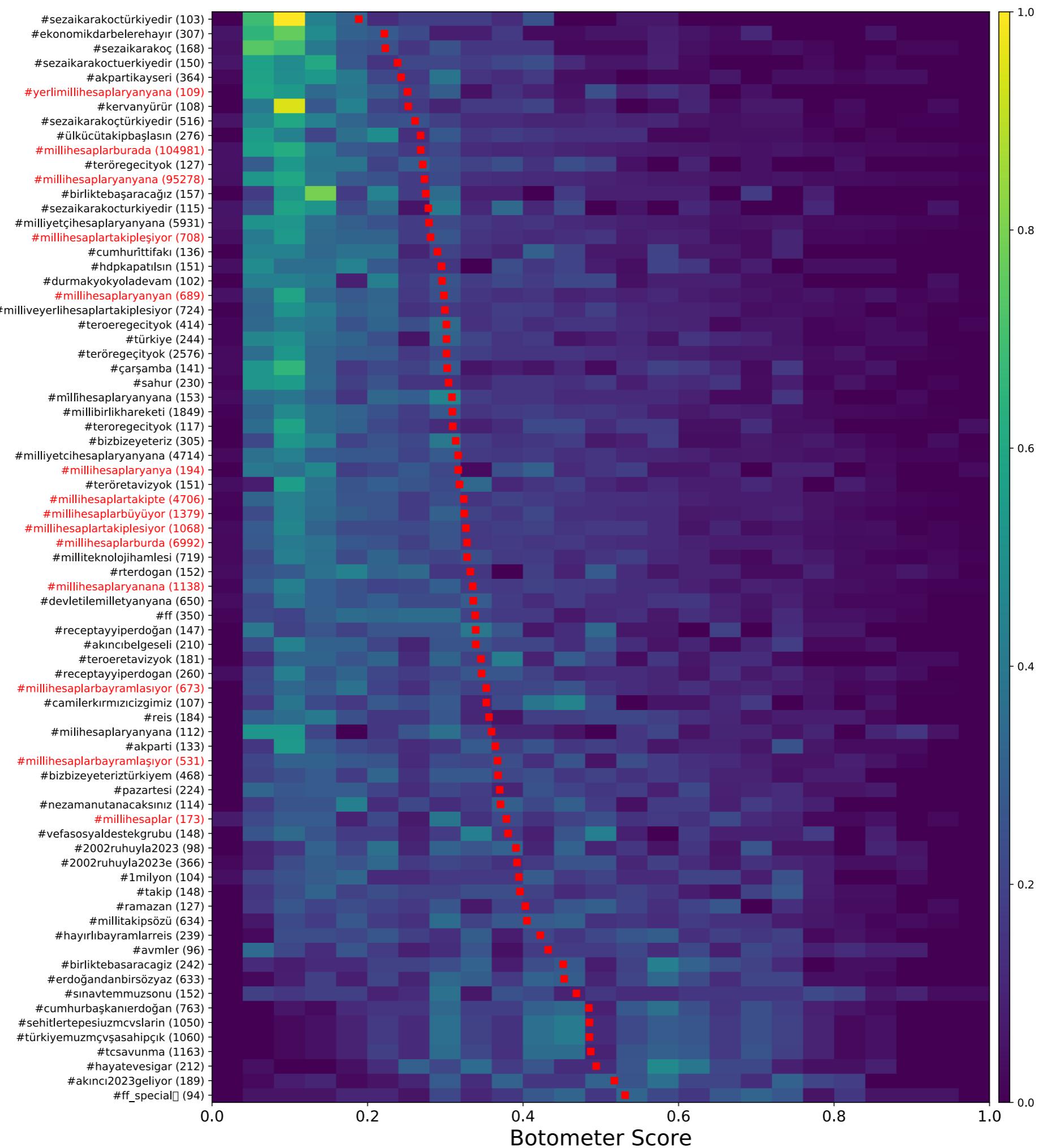
#sezaikarakoçturkiyedir #sezaikarakoc

#devletilemilleyanya  
#sezaikarakoctuerkiyedir  
#akpartikayseri

#2002ruhuyla2023e

#TÜRKİYEHESAPLARAKIPLESİYOR  
#cumhurbaşkanıerdogancumhurbaşkanı  
#tosavunmagan

#erdogandanbirşözyaz



# **Demo: Bonus curiosity driven analysis**

Let's take a look at

**90\_suspended-accounts-analysis.ipynb**

notebook for the analysis

# Thank you very much



*I will appreciate if you can spare 5 min. To fill out the survey below to help me make this tutorial better.*

*Survey link: <https://forms.gle/otnxxSAZxFq9t5bv9>*



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