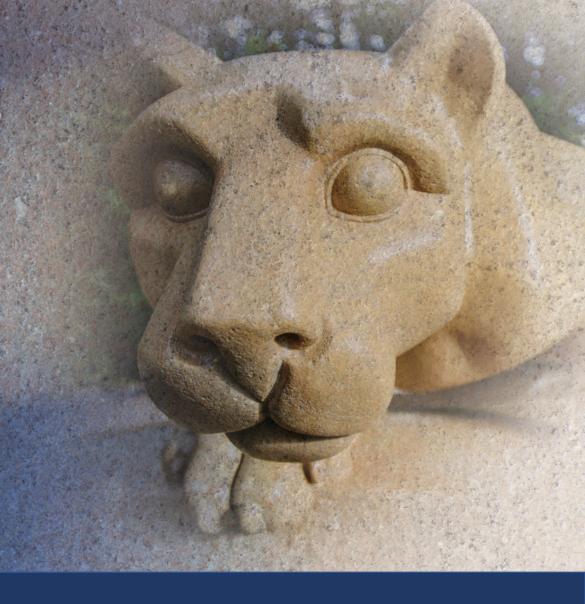
# LONGITUDINAL BEHAVIOR DYNAMICS AMONG EXTREMIST USERS IN TWITTER

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#### INTRODUCTION

- Dark side of online interactions: misinformation, trolling, cyberbullying
- Extremist groups (Al-Qaeda, white supremacists, and ISIS or the Islamic State of Iraq and Syria) have embraced social media as a tool to:
  - Recruit new members
  - Undermine their rivals
  - Spread propaganda
- ISIS has been described by the FBI as the most adept terrorist group at using Internet and social media propaganda to recruit new members.



#### **RELATED WORK**

- Alfifi et al.
  - Retweets of ISIS are more than 3 times likely to be from suspended accounts
  - So, ISIS re-tweeters are also involved in malicious activities (and are most likely ISIS supporters)
- Badawi et al.
  - Explored how ISIS makes use of social media to spread its propaganda and to recruit militants
  - Provide evidence of important offline events that are strongly intertwined with the online conversation
  - Certain topics dominating the conversation right before or after ISIS's activity spikes
- Berger and Morgan
  - Studied how suspensions were limiting the social network of English-speaking ISIS supporters
  - Individual users who repeatedly created new accounts after being suspended suffered devastating reductions in their follower counts.



### **MOTIVATION**

- Research gap in large scale analysis of ISIS reach and impact.
- Can we identify strategy/behavioral pattern that ISIS uses in Twitter?
- How does ISIS use different interaction strategies on Twitter to create a thriving online extremist community?

## **DATASET BREAKDOWN**

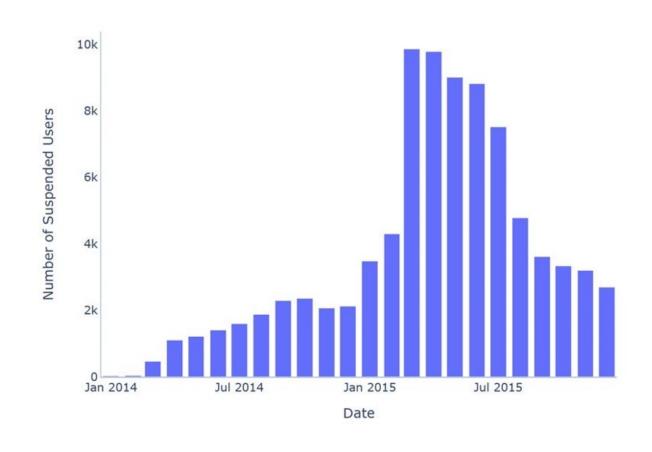
- Original dataset collected by Texas
   A&M contains 17B+ Arabic tweets
- A sample of the original dataset
  - 2,438,794 tweets from 2014 to 2015
  - ISIS and ISIS-related

Type of Tweet	Percentage
Tweets from ISIS	60.8%
Mentions of ISIS accounts	20.5%
Retweets of ISIS tweets	15.3%
Quotes of ISIS tweets	2.8%
Quotes of Mentions of ISIS	0.6%



## **USER SUSPENSIONS**

- Initially 93,159 users
- 25,000+ ISIS sympathizers (ISIS seed accounts)
  - Identified by crowdsourcing initiated by the Anonymous hacking group
  - Expanded by researchers
- Mostly got suspended by Twitter over time
- The number of users suspensions begins to drastically increase around March–July 2015





## **USER BEHAVIOR MODEL**

- We model each user as a strategic agent who will choose to:
  - Tweet
  - Retweet
  - · Quote or Mention
- 105 epochs (105 weeks during 2014-2015: 2\*365 / 7)
- Each user has 315 (= 3 x 105) features
- This can be modeled by a two-dimensional unit simplex embedded in  $R^3$  ( $\Delta^3$ )

$$\mathbf{n}^i(e) \,=\, \left(n_T^i(e), n_R^i(e), n_Q^i(e)\right)$$

$$\Delta^3 = \left\{ \mathbf{x} \in \mathbb{R}^3 : \|\mathbf{x}\|_1 = 1, \, \mathbf{x_i} \ge 0 \right\}$$

$$\|\mathbf{x}\|_1 = \sum_i |x_i|$$

$$\mathbf{x}^{i}(e) = \frac{\mathbf{n}^{i}(e)}{\|\mathbf{n}^{i}(e)\|_{1}}$$

$$\mathbf{g}(e) = \sum_{i}^{N} \mathbf{n}^{i}(e)$$

# **USER BEHAVIOR MODEL (CONT'D)**

The dataset is dominated by tweet activity.



$$\mathbf{n}^i(e) \,=\, \left(n_T^i(e), n_R^i(e), n_Q^i(e)\right)$$

$$\Delta^3 = \left\{ \mathbf{x} \in \mathbb{R}^3 : \|\mathbf{x}\|_1 = 1, \, \mathbf{x_i} \ge 0 \right\}$$

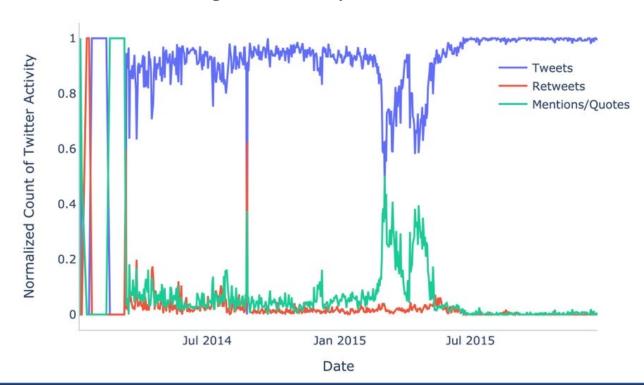
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$$\mathbf{x}^{i}(e) = \frac{\mathbf{n}^{i}(e)}{\|\mathbf{n}^{i}(e)\|_{1}}$$

$$\mathbf{g}(e) = \sum_{i}^{N} \mathbf{n}^{i}(e)$$

# **USER BEHAVIOR MODEL (CONT'D)**

- Strategy changes substantially in the middle of 2015 as a result of account suspensions.
- An increase in number of RTs and QT/MT:
  - Switch in the strategic behavior of key accounts



$$\mathbf{n}^i(e) \,=\, \left(n_T^i(e), n_R^i(e), n_Q^i(e)\right)$$

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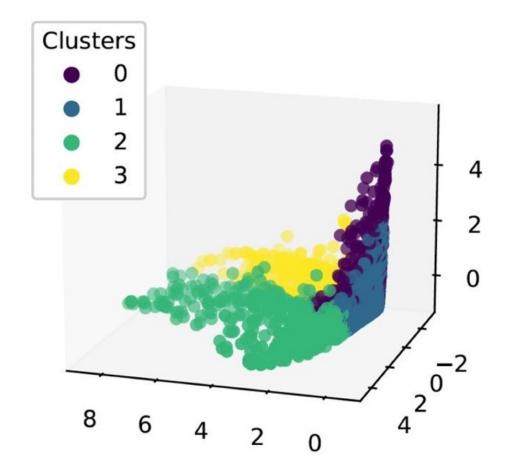
$$\mathbf{g}(e) = \sum_{i}^{N} \mathbf{n}^{i}(e)$$

$$\mathbf{x}^g = \frac{\mathbf{g}(e)}{\|\mathbf{g}(e)\|_1}$$



# **BEHAVIOR CLUSTERING**

- K-means (unsupervised) clustering
  - Best Silhouette and Davies-Bouldin scores with k=4
- GMM (Gaussian Mixture Models)
  - Best Silhouette and Davies-Bouldin scores with k=4
- K-means yielded better scores, but clusters were to some extent consistent among the two methods.
- PCA to project 3\*105-dimensional vectors into 3-dimensional vectors





#### **ROLES AND BEHAVIOR PATTERNS**

- Content Producers: tend to produce original content (tweets)
- Content Amplifiers: Use retweets, quotes, or mentions as a tool to amplify the posted tweets
- Bimodal Cluster:
  - Begin by using a mixture of tweets, retweets and quotes,
  - Then, transition to almost exclusive use of tweets
  - And then, transition back



## **CLUSTERS OF ACCOUNTS**

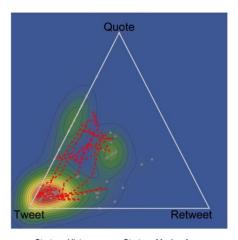
Cluster 0
Content producers

Content amplifiers

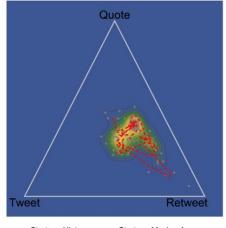
Cluster 1

Cluster 2 Bimodal Cluster 3
Content producers

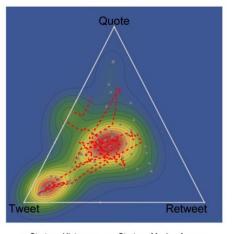
61% of 500 top retweeted users



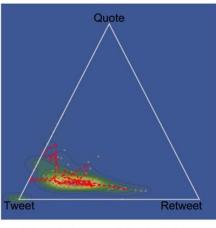
- Strategy Points



- Strategy Points



- Strategy Histogram
   Strategy Moving Average
- Strategy Points

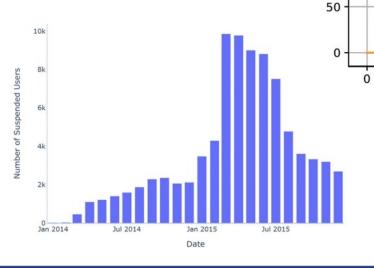


- Strategy Histogram
   Strategy Moving Average
- Strategy Points



## SHIFTED ACCOUNTS

- Number of accounts in the bimodal cluster who shifted from amplifying to tweeting and vice versa
- An increasing number of users shift from amplifying to tweeting until week 64 (2<sup>nd</sup> week of March)
- Smooth decrease in shifting from tweeting to amplifying (week 66; parts of the 1<sup>st</sup> and 2<sup>nd</sup> weeks of April)



Shift from amplifying to tweeting Shift from tweeting to amplifying

20

80

Week index out of 105 weeks

100

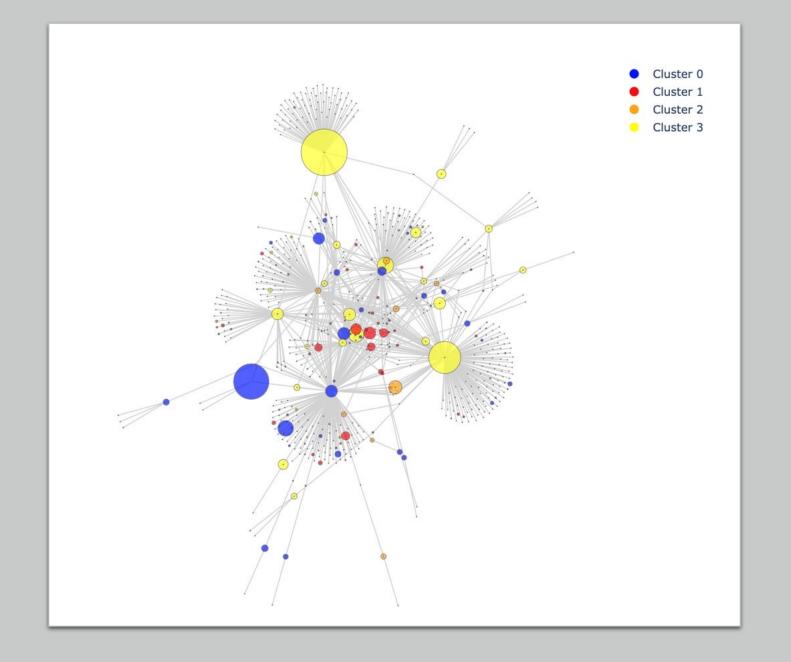
250

Number of shifted users



# RETWEET NETWORK OF TOP-50 TWEETERS

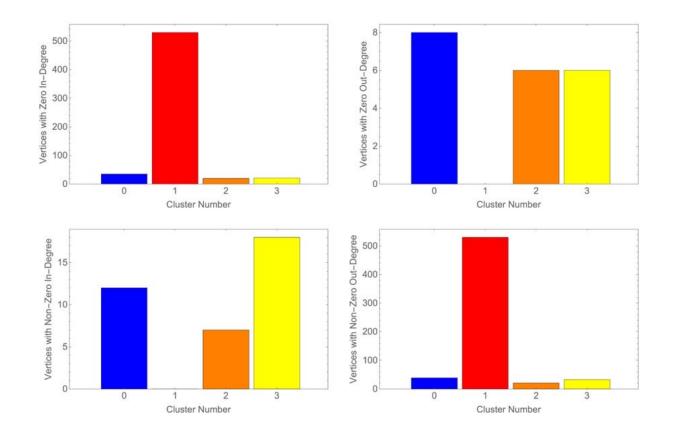
- Top 50 accounts who posted the highest number of tweets during 2014–2015
- Larger vertex = more posted tweets
- 643 vertices in total (including retweeters of top-50)
- Accounts in C<sub>0</sub> and C<sub>3</sub> (content producers) are heavily retweeted





# **CENTRALITY OF CLUSTERS**

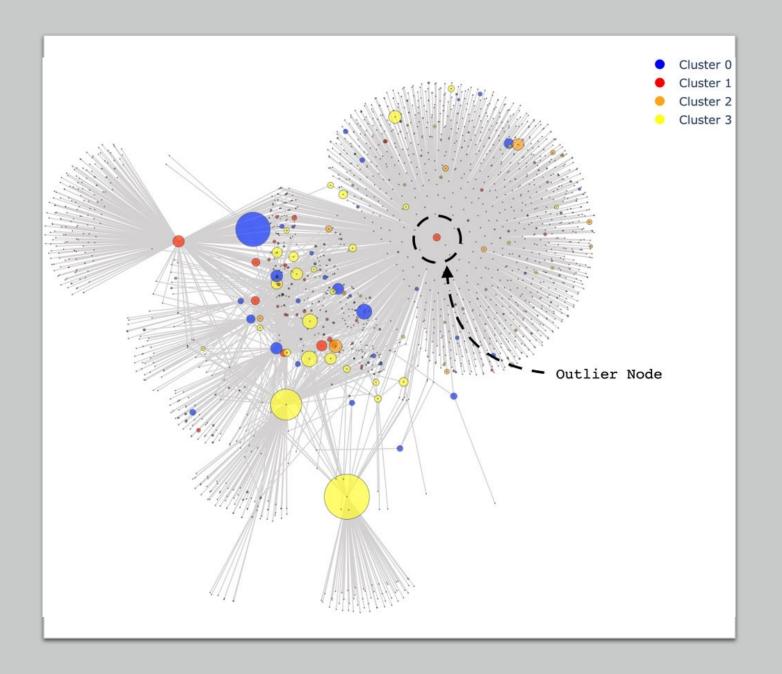
- A higher number of accounts in the amplifying cluster
   (C<sub>1</sub>) have zero in-degree (are not retweeted)
- A higher number of accounts
   from content producing (C<sub>0</sub> and C<sub>3</sub>) and the
   bimodal (C<sub>2</sub>) clusters have non-zero indegrees (are retweeted) and zero out-degrees
- Some accounts are used solely for retweeting content while other accounts are used for content production
- Supports our clustering





# RETWEET NETWORK OF TOP-50 ACTIVE ACCOUNTS

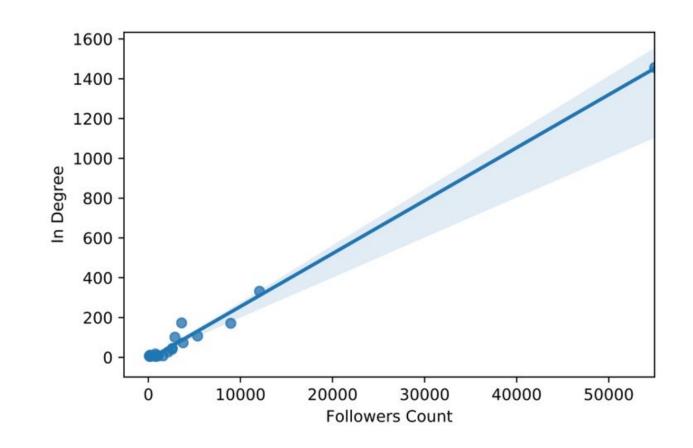
- Top 50 accounts based on total Twitter activity
- 1,927 vertices in total (including retweeters)





## **OUTLIER ACCOUNT**

- Top 20 users
- 50k+ followers
- Produced 15,000 tweets
- 64% of retweeters are non-ISIS
  - Significant impact in reaching out to accounts beyond the ISIS community

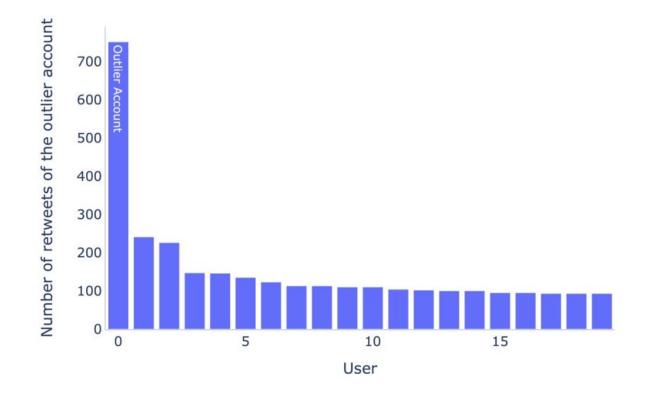




# **OUTLIER ACCOUNT (CONT'D)**

- Top 20 users
- ISIS users who retweeted the outlier account at least 100 times

Top 20 users who retweeted Outlier account





# **OUTLIER ACCOUNT (CONT'D)**

- Wilcoxon signed-rank test
  - Measure the statistical significance of difference between the behavior of the outlier and top-50 active users

Column	Result	P-Value	Stat.
Tweets	Diff.	$4.4 \times 10^{-9}$	869
Quotes/Mentions	Diff.	$2.2 \times 10^{-8}$	417
Retweets	No diff.	0.71	1622
Combined	Diff.	$3.1 \times 10^{-18}$	45

