

Viewpoint Discovery and Understanding in Social Networks

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Introduction & Motivation

- Social Media:
 - Dominant platform to comment & discuss on breaking news and events
 - Enable the expression of diverse opinions and **viewpoints**
- However:
 - Media bias
 - Echo Chamber
 - Filter bubbles
- Social costs and effects
 - As also observed in the US election 2016 (Allcott and Gentzkow, 2017)
- Need of methods to **detect** and **understand** the different viewpoints about a topic
 - To better understand political or societal debates
 - To better analyze and interpret course of historical events retrospectively



<http://theday.co.uk/technology/research-reveals-facebook-echo-chambers>

➤ Viewpoint Discovery

- Given a topic, detect the different viewpoints discussed in a Social Network (Twitter)
- Cluster the users based on their viewpoint on the topic (by exploiting retweets)

➤ Viewpoint Understanding

- What is a viewpoint about?
 - *by identifying a list of descriptive terms that characterize the viewpoint*
- What is a term about?
 - *by identifying other descriptive terms*

Related Work

- Viewpoint discovery in Documents:
 - TAM (Paul and Girju,2010)
 - VODUM (Thonet et al, 2016)
- Viewpoint discovery in Social Media:
 - SNVDM-GPU (Thonet et al, CIKM 2017)
 - Detect viewpoints of subtopics related to a controversial topic
 - Show descriptive terms that characterize a viewpoint
 - Others:
 - Supervised models [Fang et al. 2015, Cohen et al. 2013]
 - Contrastive opinion mining (Ren et al,2016)
 - Controversy detection [Garimella et al. 2016]
 - Viewpoint discovery in forums [Qiu and Jiang, 2013]

Our Approach

- Viewpoint Discovery
 - Use of **interaction graph** (retweet graph)
 - By first gathering tweets related to the topic
 - Apply **multilevel graph partitioning** (as done by Garimella et al, 2016)
 - Exploitation of **conductance** clustering quality metric
 - Find clusters representing concrete viewpoints
 - Identify noisy clusters
- Viewpoint Understanding
 - Apply a **Rank Difference** approach **iteratively**
- Contrary to existing works:
 - Applicable to unknown number of viewpoints
 - Detection of noisy groups with no clear viewpoint
 - Deep understanding of viewpoints (through focusing on specific terms)

Choosing the proper graph clustering method

- Main requirement: Ensuring sparse connection between different clusters
- No edge between the users \neq different viewpoints
 - But: users with different viewpoints will definitely not endorse each other
- Various ways of clustering
 - Node similarity
 - The Louvain method (Community Detection)
 - Multi-Level Graph Partitioning (MLGP)
 - Using METIS
 - Problem: what number of clusters to consider (**K** value)?

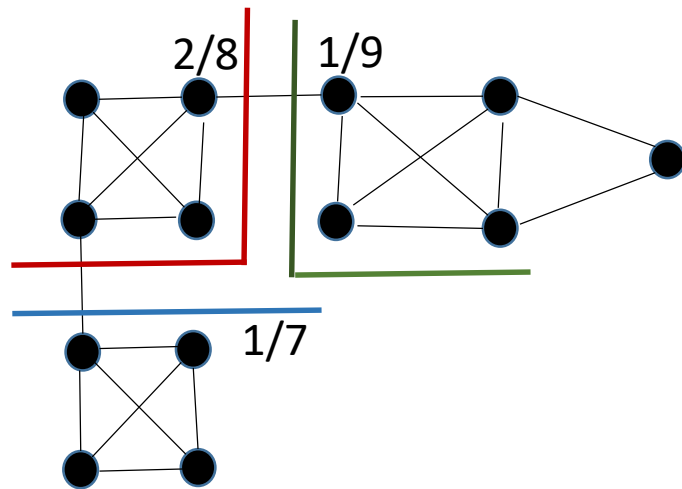
Viewpoints Discovery

➤ Conductance:

- Ensures sparse connection between the different clusters!

$$\text{Conductance} = \frac{\text{size of cut (edges connecting the clusters)}}{\text{volume of smaller cluster divided by that cut}}$$

Volume= total number of edges starting from *nodes in the cluster*



Advantages of Conductance:

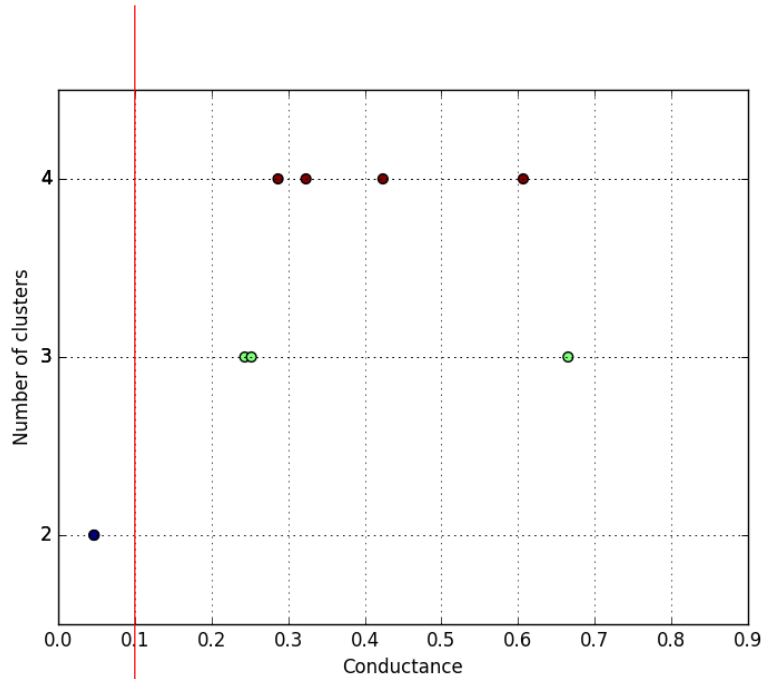
- It considers the cut size
- It considers the volume of the cluster

Deciding on the clusters that hold different viewpoints

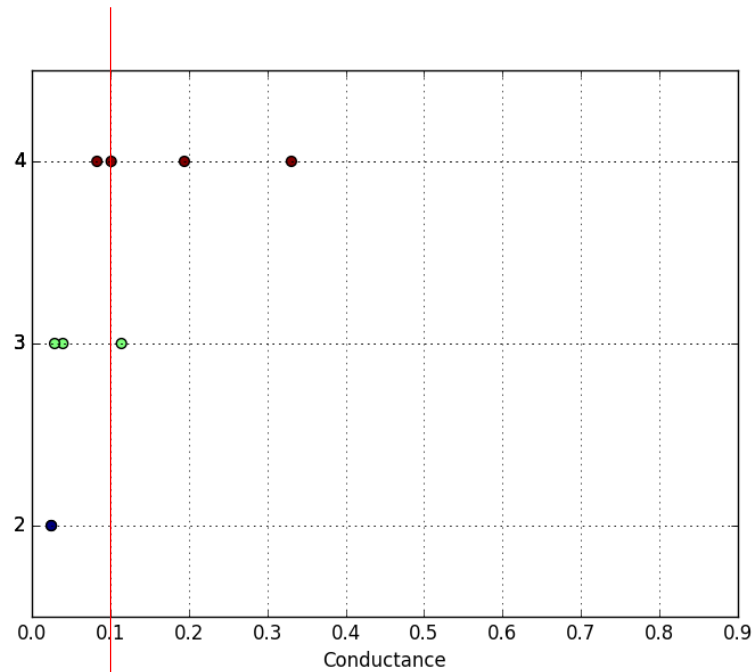
- We inspect the conductance of the clusters for different values of k
- Conductance Threshold:
 - A value of 0.1 seem to be good threshold
 - *10% of the edges in the cluster point to nodes outside the cluster!*
 - Conductance < 0.1 means well separated therefore good clustering
 - Conductance > 0.1 , means most likely noise
- For $k=2$, if the conductance > 0.1 , the topic is probably not controversial!
- Algorithm:
 - Start from $k = 2$ and check the conductance value
 - Continue for $k > 2$ until we get the maximum number of clusters below the threshold for the larger k value

Viewpoints Discovery

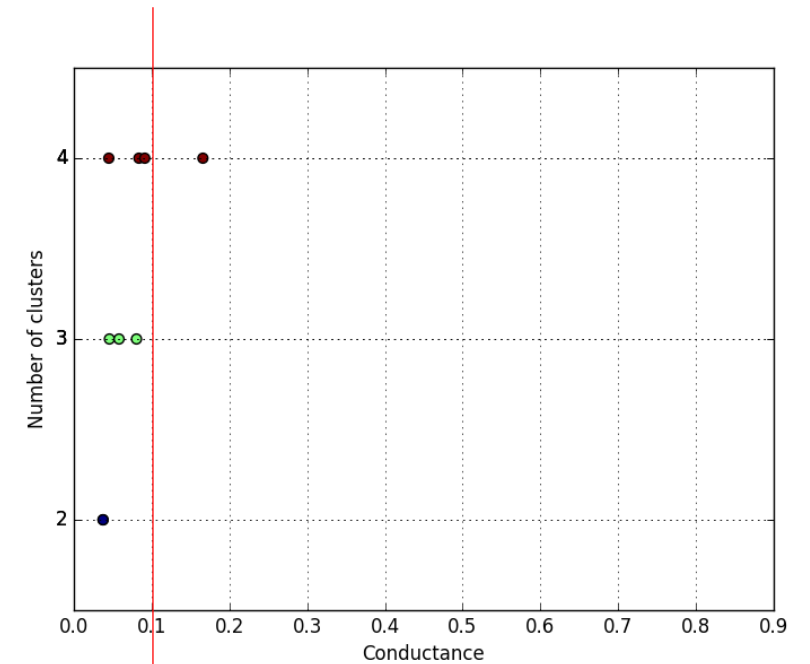
Conductance of the cut of each cluster for various k values



Indyref



US Election



Brexit

Threshold: 0.10 (Red line)

Viewpoints Understanding

- **Rank Difference:** automatic term recognition algorithm (Kit & Liu, 2008)
 - Operates over two ranked lists of terms: one extracted from the “**subject corpus**” (corpus of interest) and one extracted from a “**contrasting corpus**” (e.g., general corpus)
 - Ranking criterion for the lists: term frequency
 - Score of a term:

$$score(w) = \frac{rank(w, W_c)}{|W_c|} - \frac{rank(w, W_s)}{|W_s|}$$

- Focuses on filtering out frequent terms occurring in both lists

Indian	European
Onion	Onion
Garlic	Garlic
Chili	Black pepper
Cumin	Paprika
Ginger	Ginger
Black pepper	Tomato
Paprika	Chili

➤ Iterative Rank Difference (IRD)

1. First Iteration

- What is a viewpoint about?
- Extract Top-n terms after running Rank Difference
 - Subject list: frequent terms in tweets of a viewpoint
 - Contrasting list: frequent terms in tweets from all **the other viewpoints**

2. Next Iterations

- What is a term (w) about?
 - Subject list: frequent terms that **co-occur** with w in tweets about the viewpoint
 - Contrasting list: frequent terms that **do not co-occur** with w in tweets **about the same viewpoint**

Effectiveness of Viewpoint Discovery

- Comparative evolution with topic modeling proposed by Thonet et al. [CIKM, 2017]
- Indyref: 2 viewpoints (YES and NO)
- Midterms: 2 viewpoints (Democrat and Republican)
- Evaluation metrics:
 - Purity: proportion of users whose are correctly clustered compared with the ground truth class
 - NMI: Based on Mutual Information and Entropy

Method	Indyref		Midterms	
	Purity	NMI	Purity	NMI
SNVDM-GPU	0.969	0.800	0.964	0.778
MLGP	0.988	0.908	0.983	0.876

Purity and NMI comparison between SNVDM-GPU and Multi-Level Graph Partitioning (MLGP)

Qualitative Analysis of Viewpoint Understanding

➤ Indyref

- Ground truth dataset (YES or NO)
- First Iteration:

Viewpoint 1	Viewpoint 2
#voteyes , #yes , westminister,	#nothanks , #bettertogether ,
Meeting, independent ,	currency , #labourno , #scotdecides
#scotland , murphy, event	speech, alex, #voteno , part
national , folk	seperation

Qualitative Analysis of Viewpoint Understanding

➤ Indyref

- Ground truth dataset (YES or NO)
- Second Iteration:

murphy	{#voteyes, #yes}
leader	Scotland,
war	independent
tax	power,
labour	future,
really	country

Qualitative Analysis of Viewpoint Understanding

- US Election
 - First Iteration:

Viewpoint 1	Viewpoint 2
#nevertrump , #theresistance	#tcot , #trump2016 , #pjnet ,
#obamafarewell , #iamwithher	#trumptrain , #draintheswamp ,
#resist , #trumps , gop, tweet	video , #makeamericagreatagain ,
#gop , #trumpleaks , tax	#fakenews , breaking, god
#notmypresident , putin	#realdonaldtrump , usa, fbi

Qualitative Analysis of Viewpoint Understanding

➤ US Election

▪ Second Iteration:

#trumpleaks	fbi
#trumprussia	investigation
#impeachtrump	comey
#dworkinreport	reopen
#amjoy	director
#muslimban	case

Qualitative Analysis of Viewpoint Understanding

➤ Brexit

■ First Iteration:

Viewpoint 1(Pro-Brexit)	Viewpoint 2 (Anti-Brexit)	Viewpoint 3 (Neutral)
Democracy,	#tory, tory	#ue, unido, reino, #maga
#nexit, try, #go,	Johnson,	#stocks, #cdnpoli, europa
#referendum	#theresamay	londres, #americafirst,
#leave, #britain	Boris, #votein	royaumeuni, #trumptrain
free, control	#stopbrexit	#healthinnovations,
#leaveeu	#nhs, cost ,nhs	#banking, #pharma

Qualitative Analysis of Viewpoint Understanding

➤ Brexit

▪ Second Iteration:

control	{boris, johnson}	banking
#takecontrol,	Conservative,	#stock
Border, uncontrolled,	Biscuit,	#healthinnovations
#betteroffout,	#foreignsecraterary, #mp,	#pharma
#takecontrol,	#borisjohnson	decrease, index

Conclusion and Future Works

- Graph clustering using **multi-level graph partitioning** and **conductance** quality metric
 - Discover groups of users that represent clear viewpoints
 - Identify noisy groups of users (with no clear viewpoint)

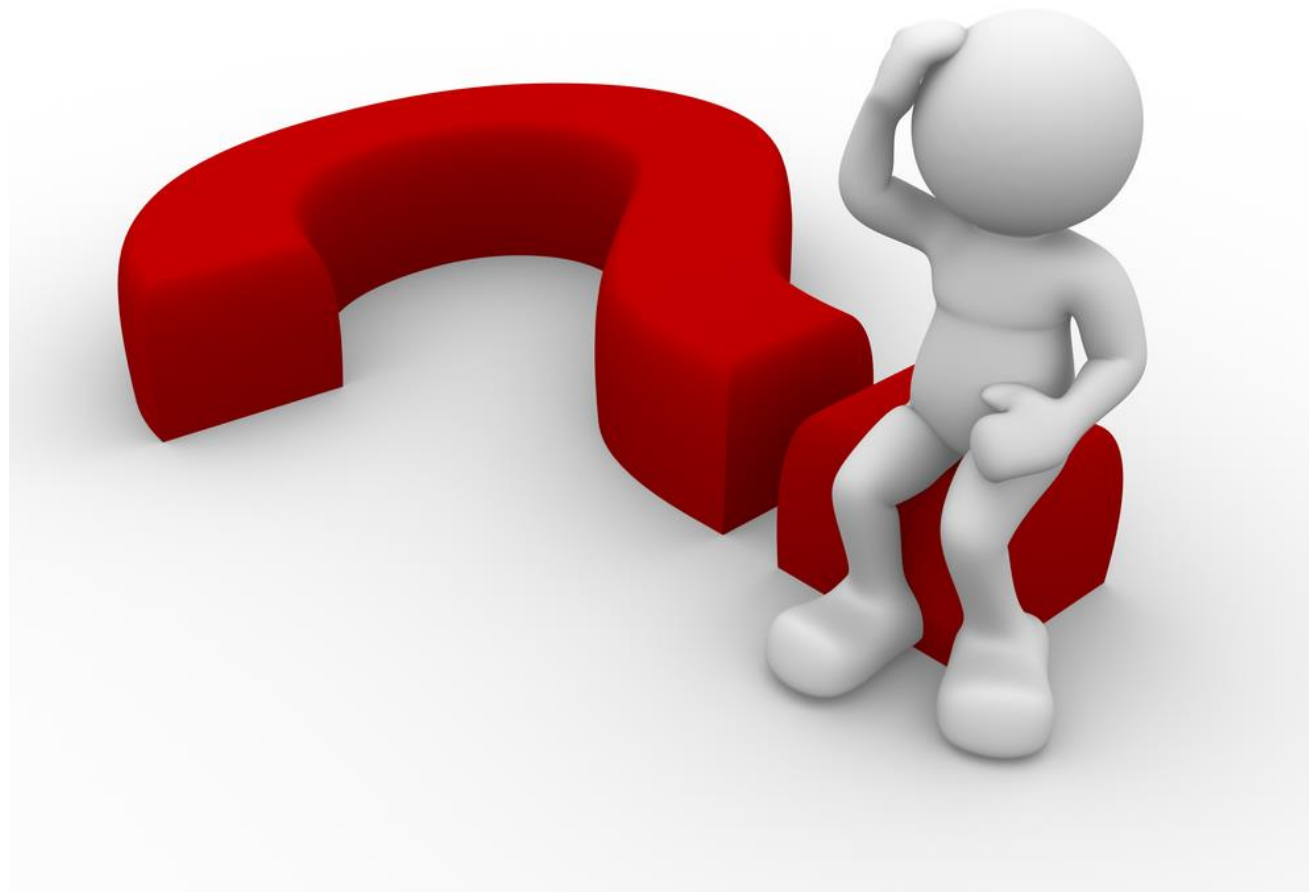
- **Iterative Rank Difference (IRD)** for understanding the viewpoints
 - 1st iteration: what is the viewpoint about
 - Next iterations: what is a particular term about?

- Evaluation
 - Proposed viewpoint discovery method outperforms more complex topic models
 - IRD allows getting a deep understanding of a viewpoint and of the related terms
 - *Why did Trump's supporters use the hashtag **#draintheswamp**?*
 - *Why did Clinton's supporters use the hashtag **#trumpleaks**?*

Future Works:

- **Timeline Summarization** of topics and viewpoints
 - that will allow understanding how a controversial topic evolves over time and with respect to the involved entities, events and subtopics
- **Knowledge graphs** about controversial topics
 - How to semantically represent topics and viewpoints to enable advanced query answering?

Thank You



Viewpoint Discovery & Understanding in Social networks, WebSci'18, Amsterdam, The Netherlands, May 27-30, 2018.