TOPIC MODELING FOR INFRASTRUCTURE-RELATED DISCUSSIONS IN ONLINE SOCIAL MEDIA

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Background

Stakeholders of Urban infrastructure projects:

- Internal (directly involved in decision making process)
 - official decision makers (Government Dept.s & Private Sectors)
 - Technical decision makers (Planners, Contractors, suppliers, etc.)

External(affected by the decisions made and related operations)

- Local Community
- General Public
- Project Affected Groups
- Local Neighborhood Members
- Pressure Groups Such As NGOs.
- News Media



Background

[External] Stakeholders Engagement:
 Public Involvement (PI Programs)

OFFLINE

Public Meetings

Open Houses

Workshops

Surveys

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ONLINE

Web 2.0

"Micro-participation"

[Evans-Cowley & Griffin 2012]



Twitter – a good example



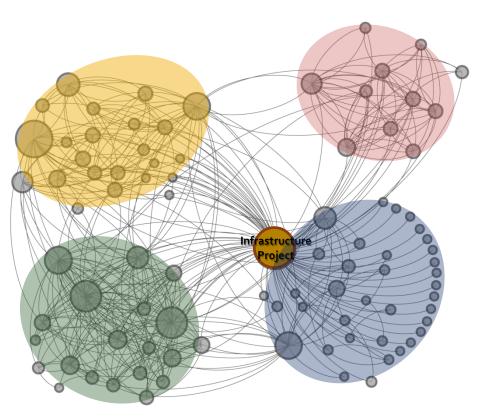
Overview

An Infrastructure Discussion Network (IDN):

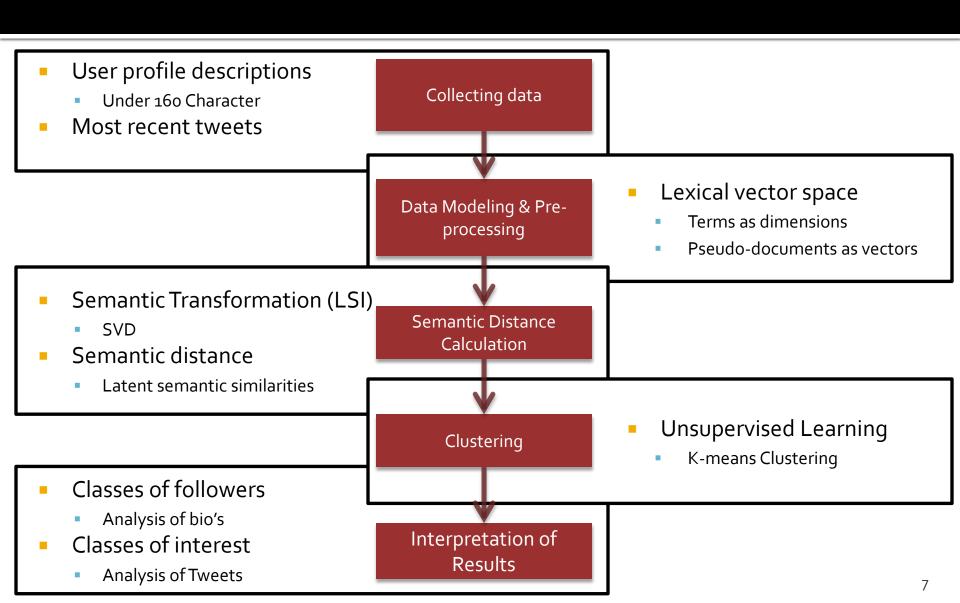


Objective

Profiling Communities & Core Interests



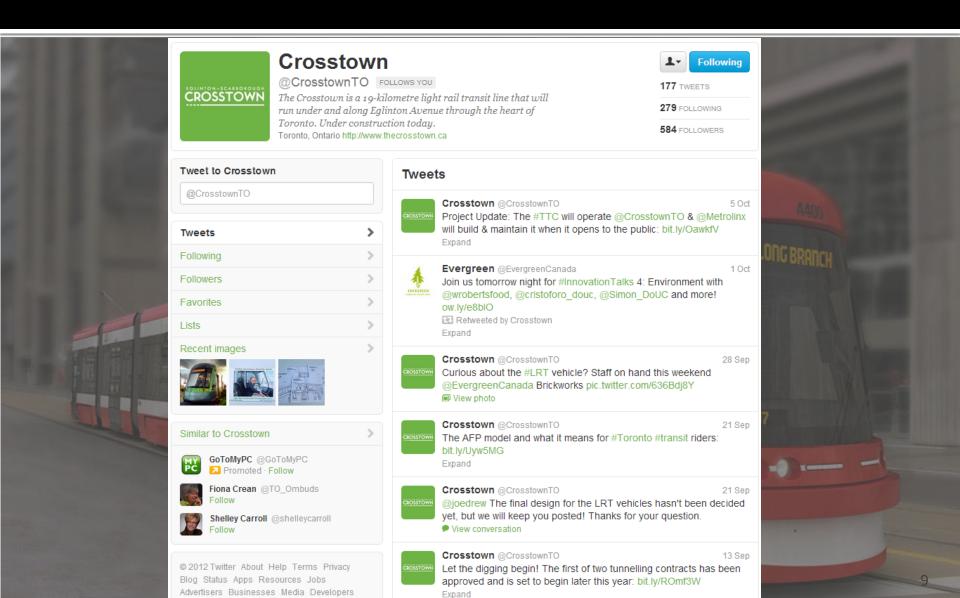
- Finding groups with
 Common Interests
 (Stakeholders Typology)
- Detecting the main Ideas discussed (Stakeholders' Vested Interests)



Case-study project

- Eglinton Cross Town LRT (TO, ON)
 CAD8.4B project
 - Connects east and west of Toronto
 - Major parts underground (10 out of 19 km)
 - Construction began in 2011
 - Now under construction

Case-study project



Modeling followers through their Profile descriptions



Killing with the control of the cont



JoshColle
Oity of Toronto Councilian Cord 15 Eglinton Lawrence
Toronto http://www.joshc.ee.co.



@JoshMatlow
Community Advocate & Toronto City Councillor for Ward 22St.Paul's. Bringing a New 2 27 (ach to City Hall.
Toronto http://www.joshmatlow.ca



Crosstown

@CrosstownTO FOLLOWS YOU

The Crosstown is a 19-kilometre light rail transit line that will run under and along Eglinton Avenue through the heart of Toronto. Under construction today.

Toronto, Ontario http://www.thecrosstown.ca



Following

177 TWEETS

279 FOLLOWING

584 FOLLOWERS



K.....

@TTCchair

ITC #OneCity governance tweets by @karenstintz & @jpboutros.For #TTP4 (a ional info/answers.follow @ttchelps @ttcchris wbraattc @ttcnotices.

Toronto http://is.gd/59mXQU



C|||||||||||d

@TTCchris

Chief Customer Officer of the Toronto Transit Commisssion.

@ttchelps for queries, com the sand complaints. @ttchelps for general for service updates.

Toronto http://www.ttc.ca



B/////////S

Toronto - http://www.ttc.ca

Executive Director, Corporate Communications, Toronto Transit Commission. Follow @tto programmers, kudos, complaints and @ttenotices for servi

Each user as a pseudo-document (psd)

- A dictionary of all (say M) terms
- Each psd as a vector:

$$\begin{array}{c} IDN \ Nodes \ (size:N) \\ \hline Dictionary \ terms \\ (size:M) \end{array} \begin{bmatrix} \overrightarrow{d(X_1)} \end{bmatrix}_{M\times 1} \ \dots \ \overrightarrow{[d(X_N)]}_{M\times N} \end{bmatrix}_{M\times N}$$

[Semantic] Similarity between vectors

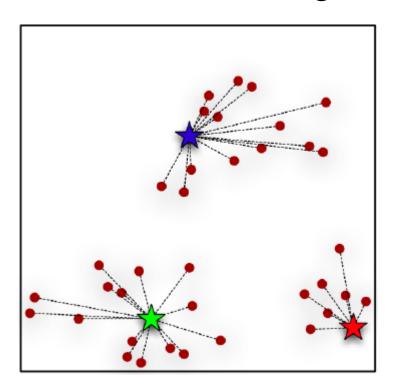
$$\mathbf{A} = \mathbf{U} \times \mathbf{S} \times \mathbf{V}^T = \widehat{\mathbf{A}}$$

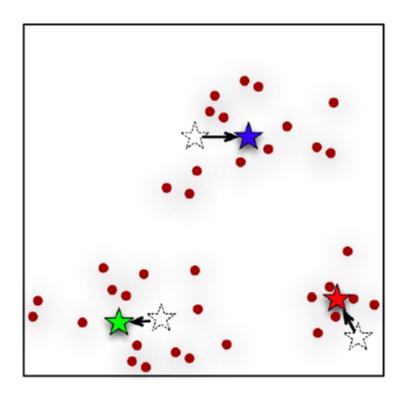




N followers (users) – N Pseudo- Documents

K-means Clustering





Selection of K:

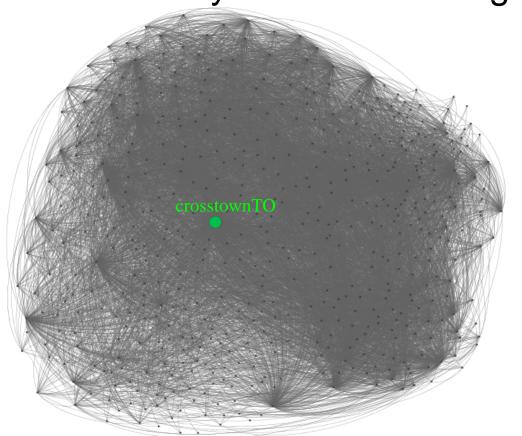
k	Clusters	Clusters sizes	Average centroid distance for each cluster	Average inter- community distance	Performance (Davies- Bouldin Index)	
2	C1	203	0.008	0.007	1.920	
	C2	204	0.006	0.007		
	C1	174	0.004	•	1.653	
3	C2	96	0.012	0.006		
	C3	137	0.005			
	C1	53	0.012		1.425	
4	C2	78	0.006	0.005		
	C3	119	0.004	0.003		
	C4	157	0.004			
	C1	110	0.004		1.730	
	C2	72	0.005			
5	C3	84	0.004	0.008		
	C4	97	0.003			
	C5	44	0.012			
6	C1	73	0.005		1.579	
	C2	91	0.004			
	C3	91	0.003	0.007		
0	C4	45	0.012	0.007		
	C5	7	0.000			
	C6	100	0.004			

Results (Main groups of followers)

	Cluster1	Cluster2	Cluster3	Cluster4
Size	157	119	78	53
	fan	transit	city	culture
	enthusiast	York	news	interested
	junkie	Lrt	civic	transit
	coffee	Mo`m	account	cities
	nerd	Work	world	housing
	political	TTC	awesome	enthusiast
	beer	metrolinx	business	photographer
	father	updates	mayor	design
	music	association	Town_hall	music
	views	Area	ward	planner
	member	village	senior	transport
	cyclist	good	twitter	issues
	news	enjoy	things	economics
	writer	project	active	interests
	avid	theatre	citizen	place
	addict	construction	resident	fan
	designer	leaside	affairs	avid
	technology	works	district	baseball
	grad	transport	job	human
	runner	environment	advisory	pop_culture
	resident	chair	progressive	policy

Results (Main groups of followers) - Comparison

Community-detection through SNA (bottom-up)



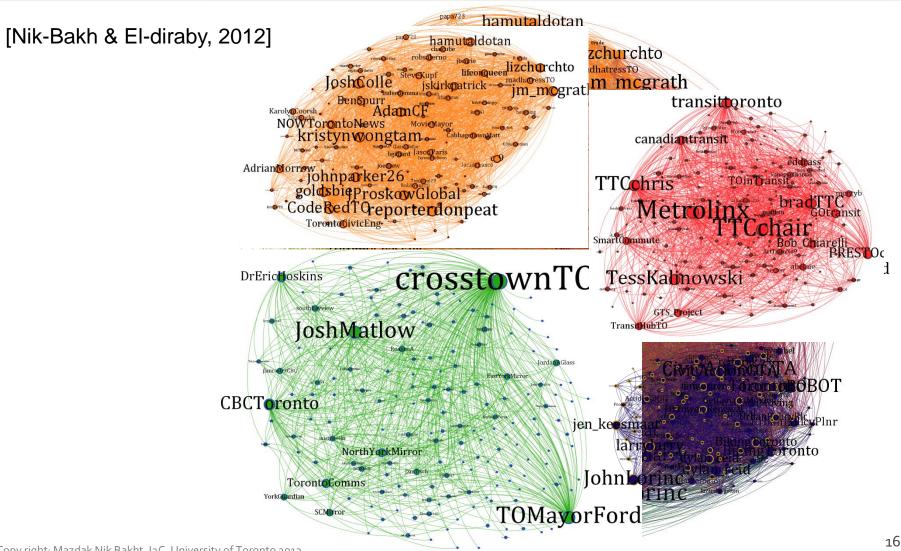
[Nik-Bakh & El-diraby 2012~13]

Each Node shows:
One project follower

Each Edge is the notion of a form of friendship/followership:



Results (Main groups of followers) – Comparison



Results (Main groups of followers) - Comparison

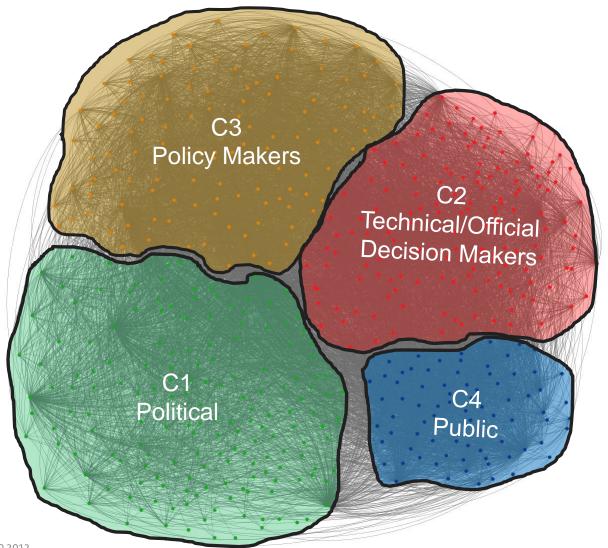
[Nik-Bakh & El-diraby, 2013]

C1	C2	C 3	C4
university	go	waste	journalist
eglinton	public	comms	cityhall
davisville	area	energy	freelance
indusry	construction	strategist	torontoist
town	hamilton	sustainable	author
updates	dedicated	commuter	civic
air	#ttc	economics	magazine
adventure	gta	ryersonu	write
business	#ttchelps	ceo	neighborhood
tv	#ttcnotices	councillor	columnist

Analysis of Bio's in communities detected through SNA

Results (Main groups of followers)

Results of Top-down & Bottom-up Analyses Coincide!



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Results (Main ideas discussed)

Economy

 XXX (representing monetary values), Dollars, Dollars_Risk, Investing, Millions, Oaarchitects, Construction_Safety, Mega_Contract

Technical features

 Construction, Safety, Contract, Consultant, Contractor, Procurement, Procurement_Process, LRT_Project, and Tunnel_Boring

Community related issues, local/regional aspects

 Neighbourhood, Community, Neighbourhood_Revitalization, Eglinton, Yonge_Bus, Finch, and Finch_LRT, Realestate

Political issues

Councilor, City_Staff, Provincial_Government, Harper

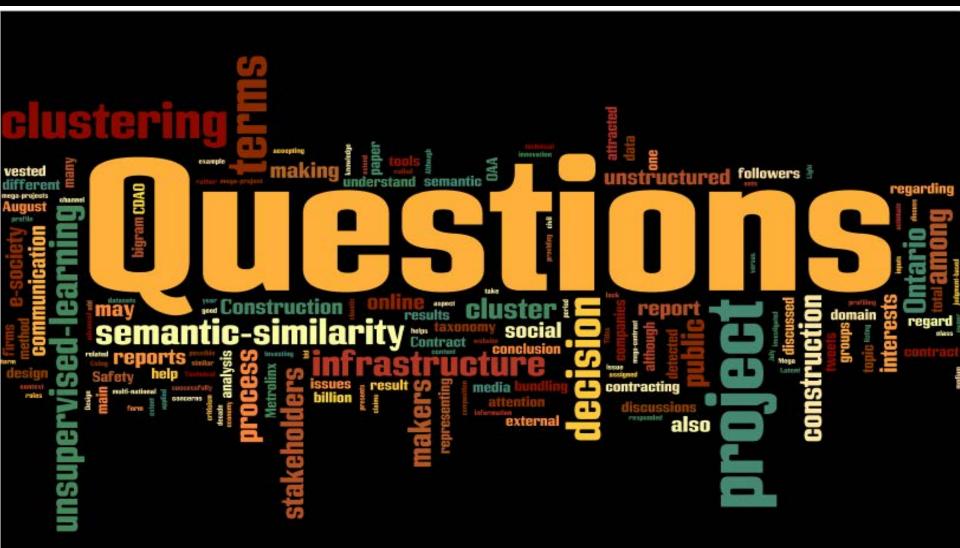
A small cluster with no specific theme

 Manual screening: News and updates about the project, its schedule and improvement, and public meetings

Conclusion

- Stakeholders typology
 - Bottom-up (Through SNA)
 - Top-down (Through semantic clustering)
- Topic detection
 - Subjectivity of interpretation
 - Existing anomalies
- Future work
 - Semantic distance through an external knowledgebase/taxonomy
 - Higher levels of automation(?)

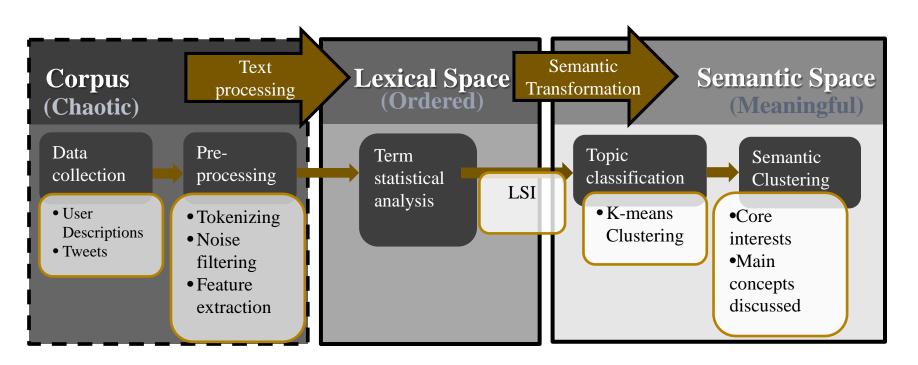
THANKS – Questions?



Literature

Author	Data-set Domain	Goal	Features	Feature values	Distance measure
Kennedy et al. (2007)	Flickr	Detection of events and places from clusters of social annotation	Terms	TF-IDF	Geographic distance
Farahat & Kamel (2010)	Various benchmark data-sets	Document clustering	Terms with stemming	TF-IDF	Variants of corpus-based semantic similarity
Antai et al. (2011)	Pdf text files	Subject classification and topic detection	Terms		LSI semantic distance
Rangrej et al. (2011)	Twitter	Clustering tweets	Terms	TF-IDF	Cosine similarity and Jaccard coefficient
Xu & Orad, (2012)	Twitter	Topical clustering and detecting representativeness	Wikipedia terms	TF-IDF	Semantic distance through an external knowledgebase (Wikipedia)
Muntean et al. (2012)	Twitter	Semantics behind hashtags	Terms	TF-IDF	Jaccard distance measure
Kireyev et al. (2009)	Twitter	Crisis and disaster detection	Domain and event- specific terms	Bag of words	Cosine similarity
Avanija & Ramar (2013)	Usenet group dataset	Web documents clustering	Terms	TF-IDF	Semantic distance over an external ontology
Suganya & Srinivasan (2013)	Web search queries	To infer query's goal from feedback sessions	Pseudo Terms	TF-IDF	Semantic distance over an external knowledge-base (WordNet)

Analysis



K-means Clustering

