

## Community-scale Online Network Analysis of Social Data Streams



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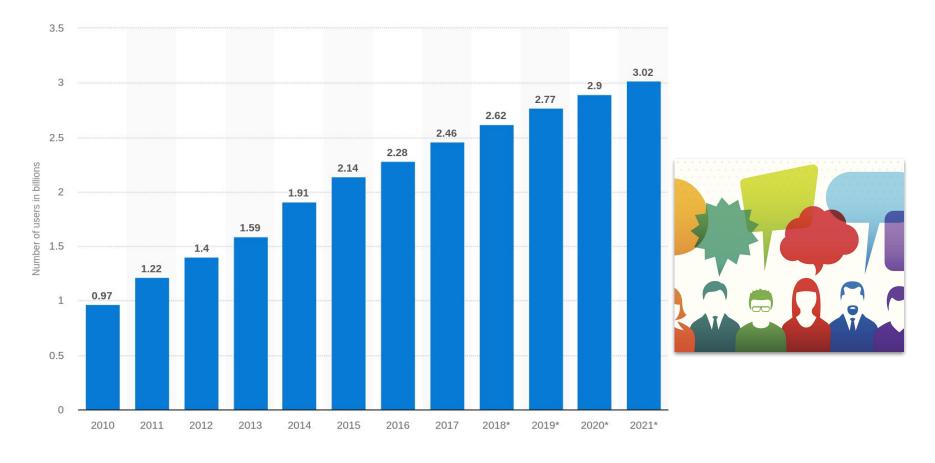
#### Outline

- Introduction & Motivation
- Methodology
- Findings and Discussion
- Conclusion

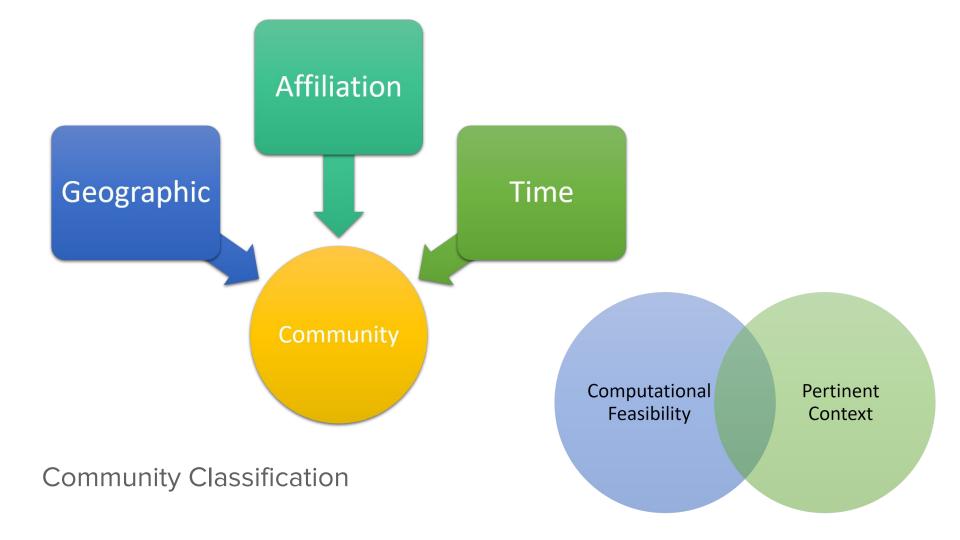
# Introduction & Motivation

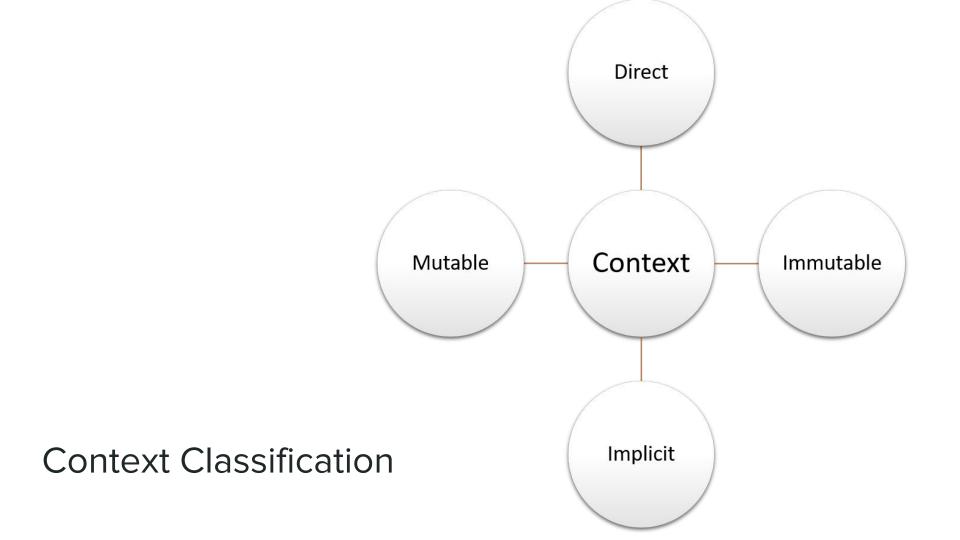


China's Social Credit system

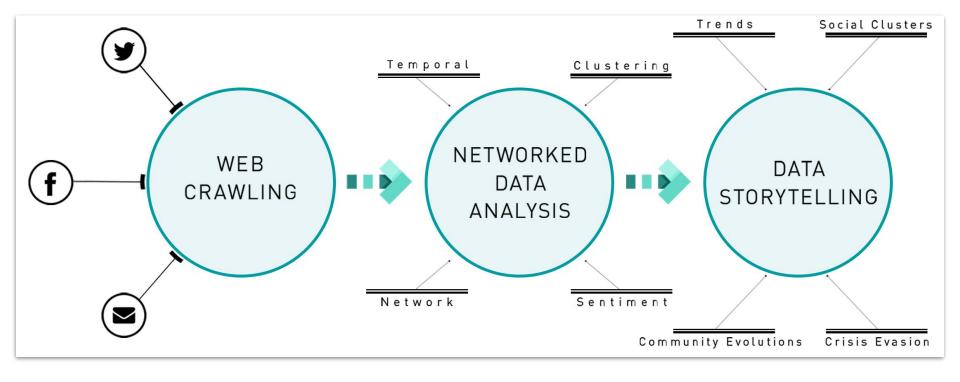


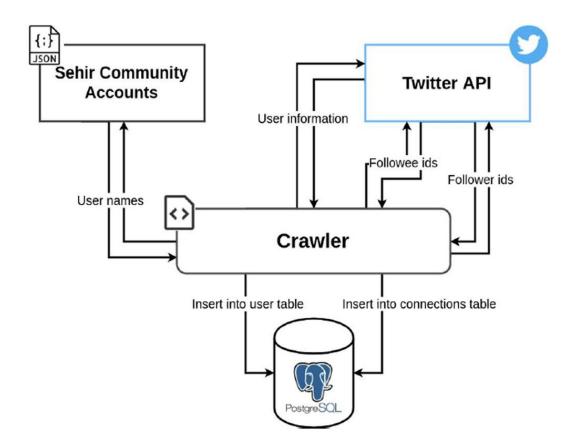
Infeasibility: Computational & Conceptual





### Methodology





Web Crawling

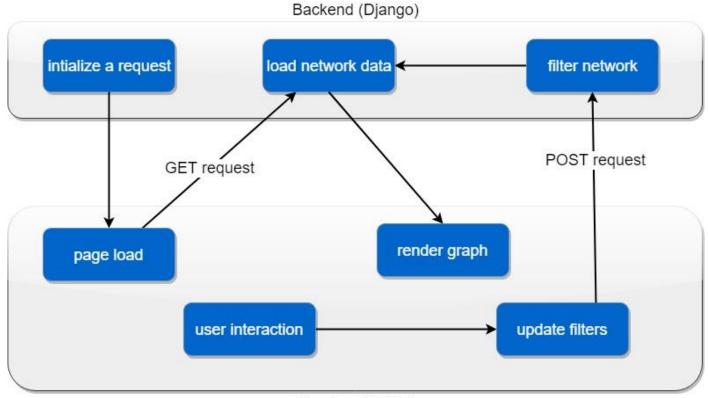
	Affiliation Netw.	Users Netw.
Nodes	1041	554
Edges	2286	951
Degree	4.39	3.43
Betweenness	$6.6e{-4}$	$9.6e{-4}$
Pagerank	$9.6e{-4}$	$1.8e{-3}$
Closeness Cent.	0.0912	0.0617 0.014 0.113
Eigenvector Cent.	$8.7e{-3}$	
Clustering Coeff.	0.306	
Transitivity	$5.14e{-2}$	455-2
Modularity	0.4535	0.619
Diameter	7	8

Table 1: Network state on 24/05/2018

#### Static Network Analysis

	CMER	HFN	Homophily
Affiliation Netw.	0.48250	0.38277	Weak
Users Netw.	0.51735	0.39109	Weak

Table 2: *Homophily* in the network

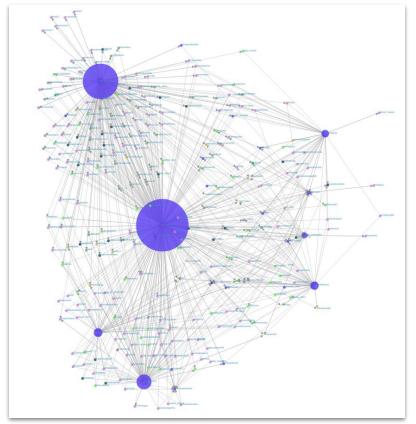


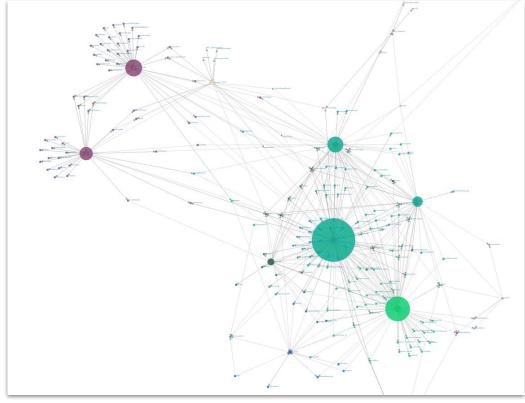
Frontend (D3.js)

#### Data Storytelling

Find Node Reset Date: 2018.05.24 Recalculate node metrics: degree: 2 7.254098360655738 100 betweenness: 0.0 0.00535429639999933 0.0 1.0 pagerank: 0.0 0.0040983606557377025 0.0 1.0 closeness: 0.0 0.22855793585206302 0.0 1.0 eigenvector: 0.0 0.028431317400879667 0.0 1.0 Clustering Coefficient: 0.0 0.7219130650905823 0.0 1.0 Nodes size by: ● degree ○ in\_degree ○ out\_degree ○ betweenness ○ closeness\_centrali filter

#### **Network Filters**





Network Screenshot

Findings and Discussion

#### Conclusion