An Introduction to Social Network Analysis With Gephi

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Do More With Digital Scholarship Series







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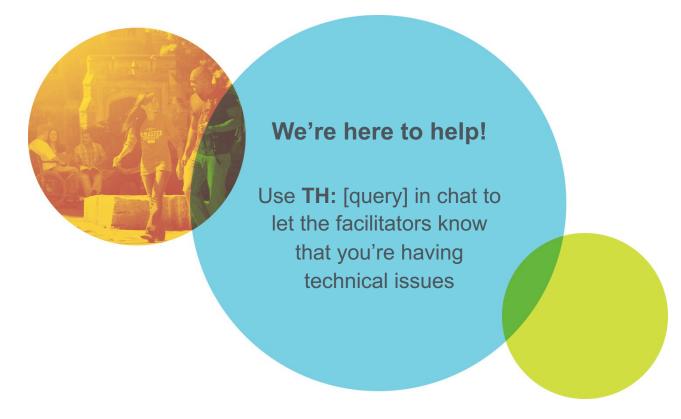
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You'll be able to:

- Define key concepts in social network analysis
- Explain what phenomena can be observed through social network analysis
- Create a network visualization in Gephi







Have you downloaded Gephi?

Go right to the source: https://gephi.org







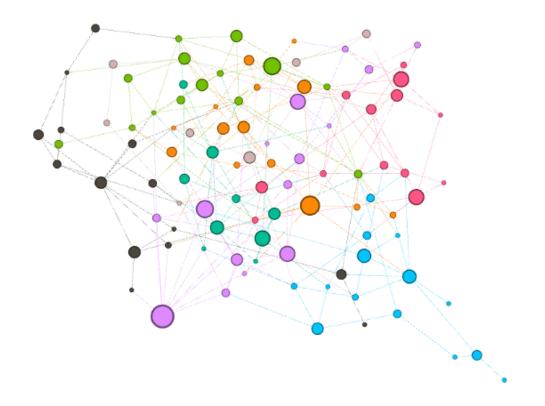
This is not a comprehensive course on social network analysis

Social network analysis is a methodological approach to representing the shape and characteristics of social structures.

 i.e. visualizing relationships between interdependent entities









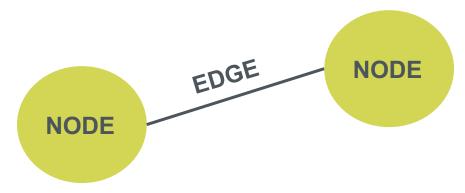




Node: the 'actor' in the network

Edge: the relationship connecting actors

Attribute: features of the node or edge









Edges can be **directed** or **undirected**.

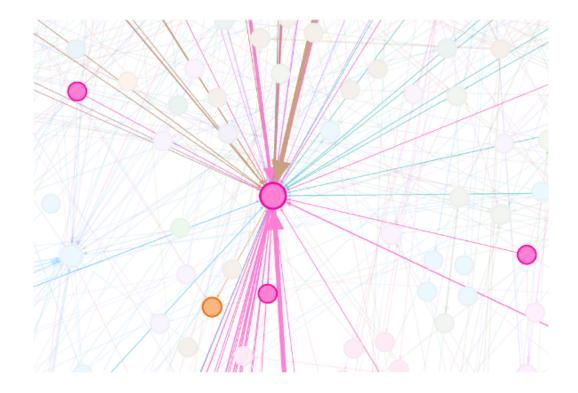
Directed edges allow us to calculate **in-degree** and **out-degree**.

- In-degree: number of incoming directed edges
- Out-degree: number of outgoing directed edges

Directed edges can also have a relative weight.















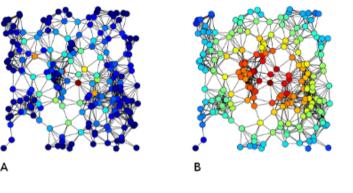
Determining which nodes are the most important in the cluster or graph...

- Degree
 - nodes with the most connections (i.e. edges)
- Closeness
 - o nodes closest to all other nodes (as a path)
- Betweenness
 - nodes which bridge the shortest paths
- Eigenvector
 - o nodes that have a higher relative influence



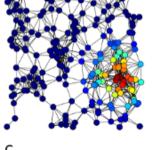


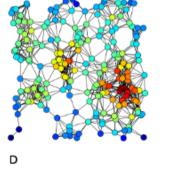
A. Betweenness



B. Closeness

C. Eigenvector





D. Degree

Image: <u>Tapiocozzo</u> (CC 4.0 BY-SA)





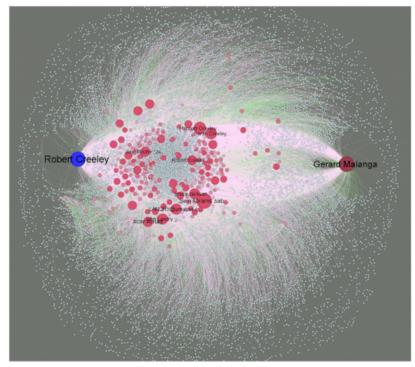


Many applications in digital scholarship:

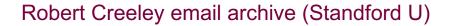
- Words that appear together often in a text (i.e. concordance)
- Correspondence between people (e.g. email or twitter)
- Communities in social networks (e.g. facebook friends)
- etc.







https://dhs.stanford.edu/visualization/robert-creeley-e-mail-correspodence-network/







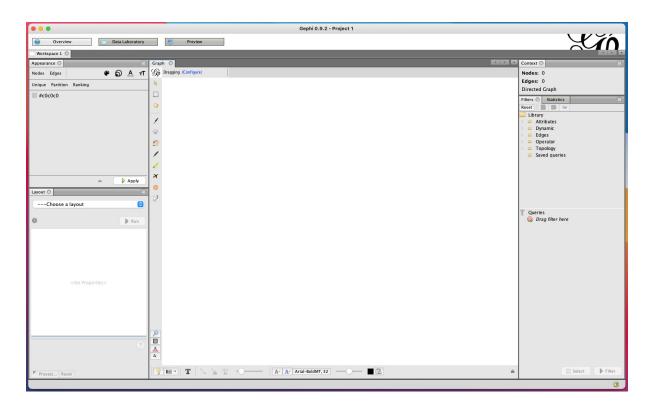


Download the #elxn43 debate night dataset

- Scraped from Twitter on Oct 7, 2019 → use of #elxn43 hashtag
- Pre-processed in OpenRefine
 - Isolated original tweets (excluded retweets)
 - Extracted mentions (@) from tweet text
 - Created source-target row for each mention



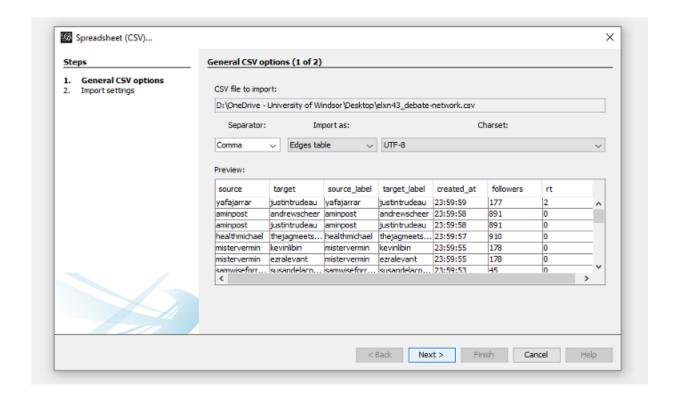








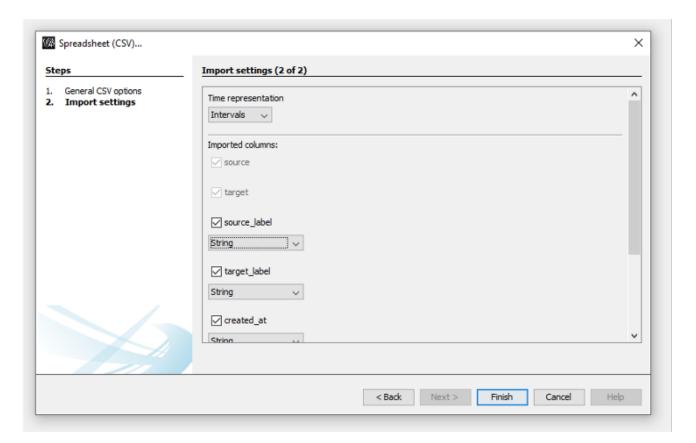






Import dataset (CSV) as: Edges table

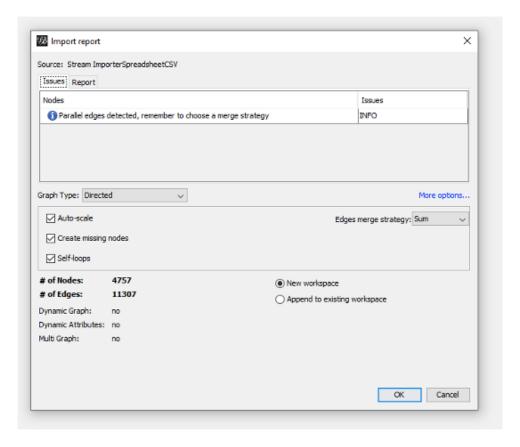




Note: name columns "source" and "target"







Almost there... directed graph?



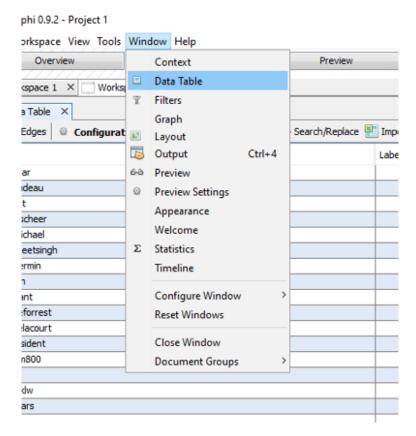








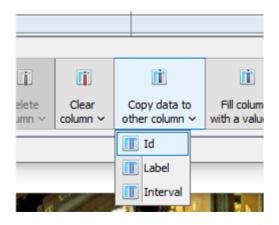












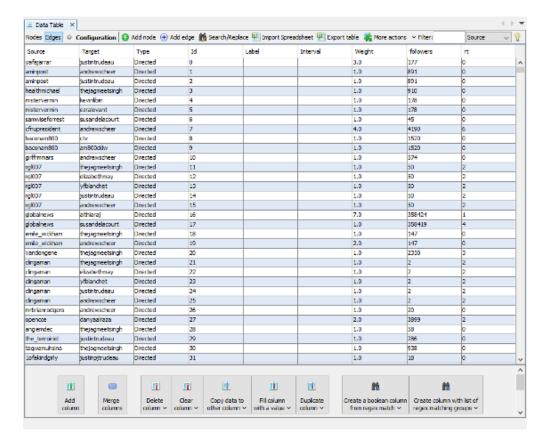




Copy data from Id to Label (in Node view)



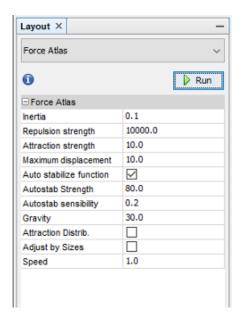


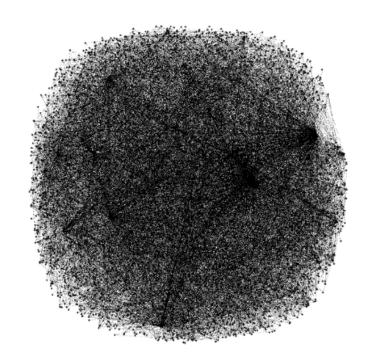


What does 'weight' refer to...?





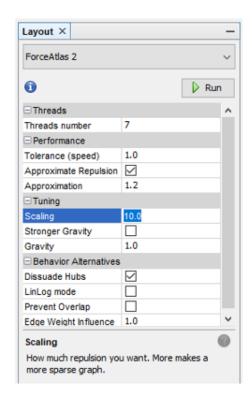


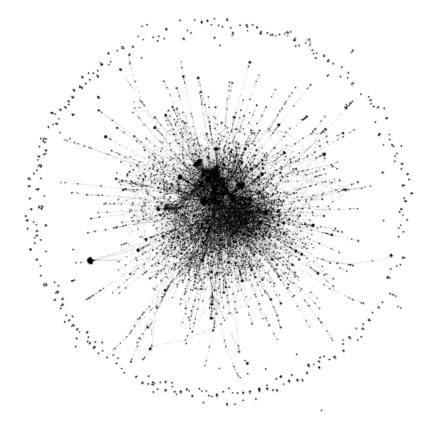


Layout (begin with Force Atlas)





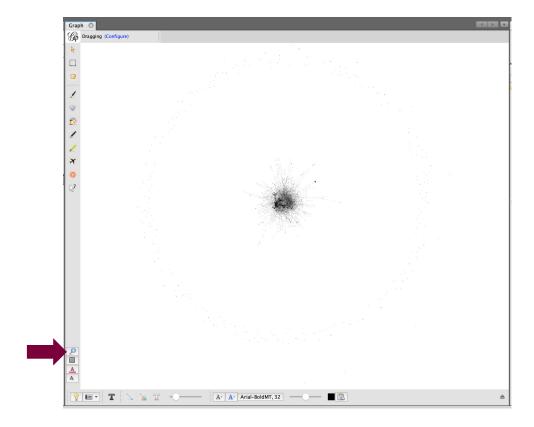




Try ForceAtlas 2...



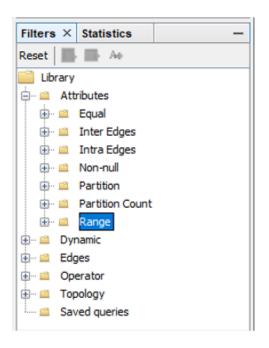


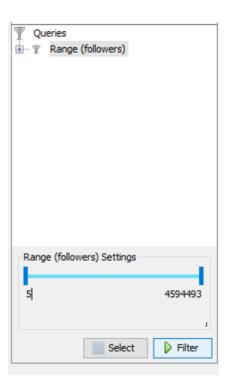








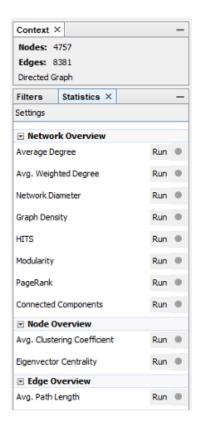




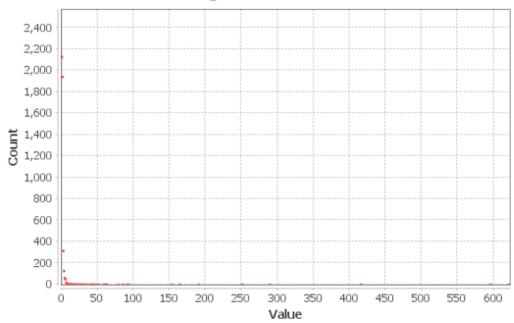
Filters to omit nodes (e.g. bot accounts)







In-Degree Distribution



Run **Statistics** for more attributes...



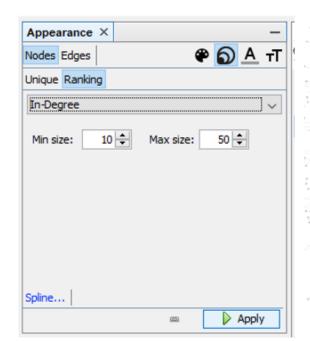


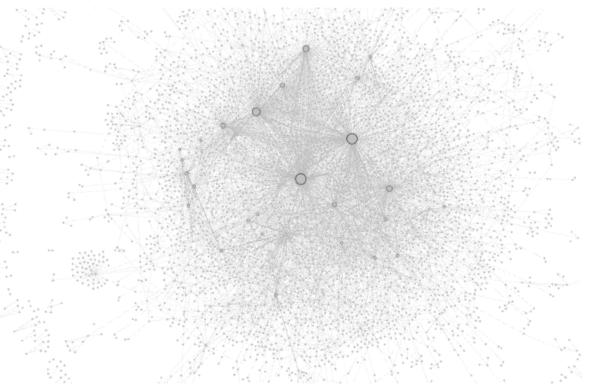
Id	Label	Interval	In-Degree	Out-Degree	Degree
justintrudeau	justintrudeau		621	1	622
andrewscheer	andrewscheer		595	0	595
thejagmeets	thejagmeets		416	0	416
elizabethmay	elizabethmay		289	0	289
cpc_hq	cpc_hq		251	3	254
maximebernier	maximebernier		190	0	190
yfblanchet	yfblanchet		164	0	164
liberal_party	liberal_party		153	1	154
althiaraj	althiaraj		93	0	93
ndp	ndp		91	1	92
gmbutts	gmbutts		85	4	89
fordnation	fordnation		78	0	78
canadiangre	canadiangre		63	1	64
cbcnews	cbcnews		62	1	63
lisalaflammectv	lisalaflammectv		61	0	61
susandelaco	susandelaco		59	2	61
rosiebarton	rosiebarton		58	0	58
torontostar	torontostar		51	0	51
jkenney	jkenney		51	0	51
globalnews	globalnews		47	5	52
davidakin	davidakin		45	10	55
ctvnews	ctvnews		43	7	50
peoplespca	peoplespca		42	0	42

Statistics are also populated in the Data Table





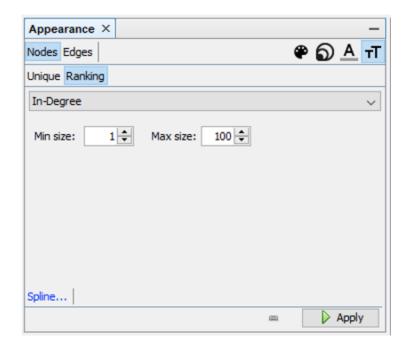


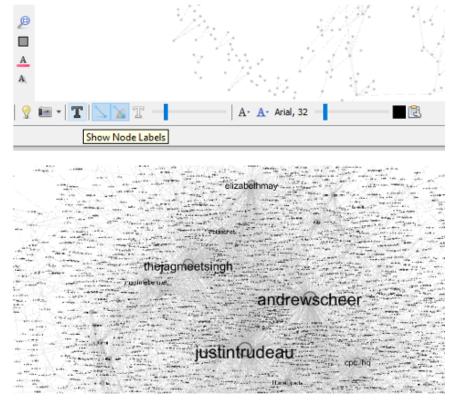










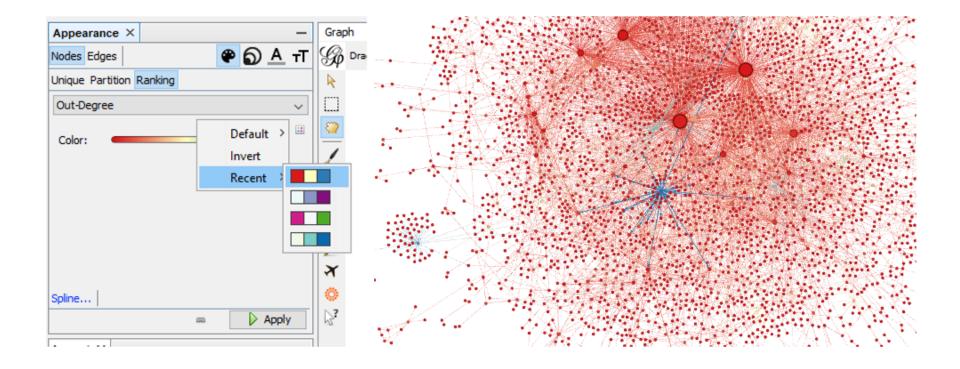


Adjust label size to identify account names





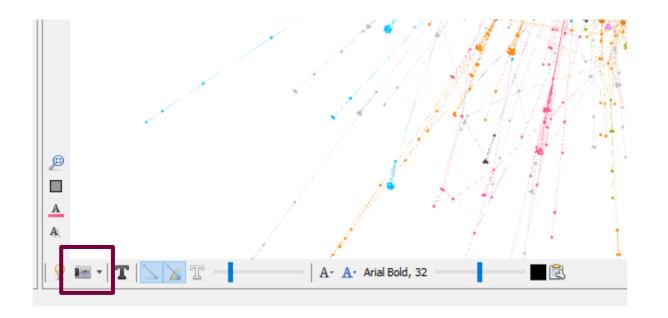
Library





Who is doing most of the tweeting?

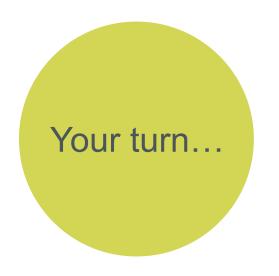






TAKE A SCREENSHOT... recall provenance



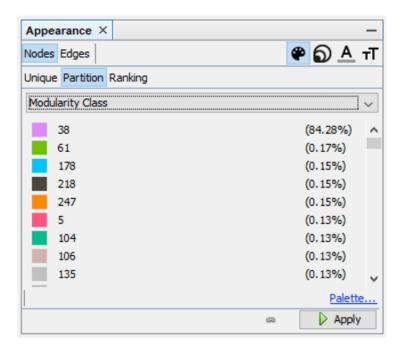


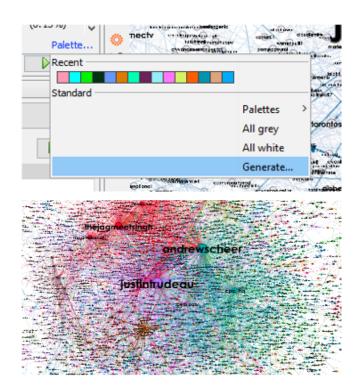
Start making sense of the data

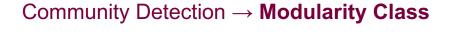
- Use attributes...
 - Calculated: In-degree, out-degree, modularity, etc.
 - Supplied: followers, retweets
- ...to modify:
 - Node & edge colour
 - Node size
 - Label size & colour





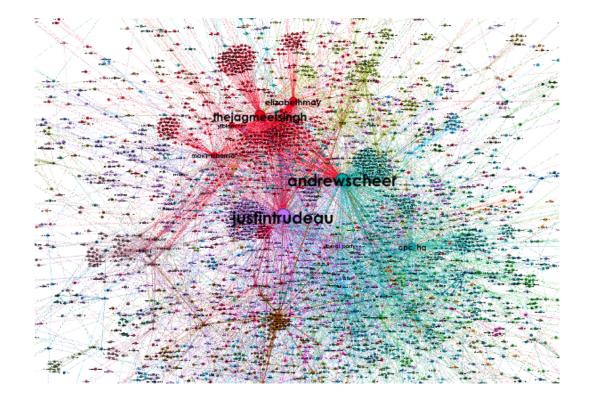




















Export as graphic to preserve layout (consider privacy, though!)

- PDF
- PNG
- SVG





Thanks for coming!

Questions: mordelld@mcmaster.ca



