First course in Network Science Sections 4.3-4.5

The content of this presentation is based on the book: A First Course in NETWORK SCIENCE. Filippo Menczer, Santo Fortunato, Clayton A. Davis, ISBN: 9781108471138, Cambridge University Press.

Sarah Alidoost Analytics SIG, 15/03/2021

Filippo Menczer, Santo Fortunato and Clayton A. Davis A First Course in NETWORK SCIENCE

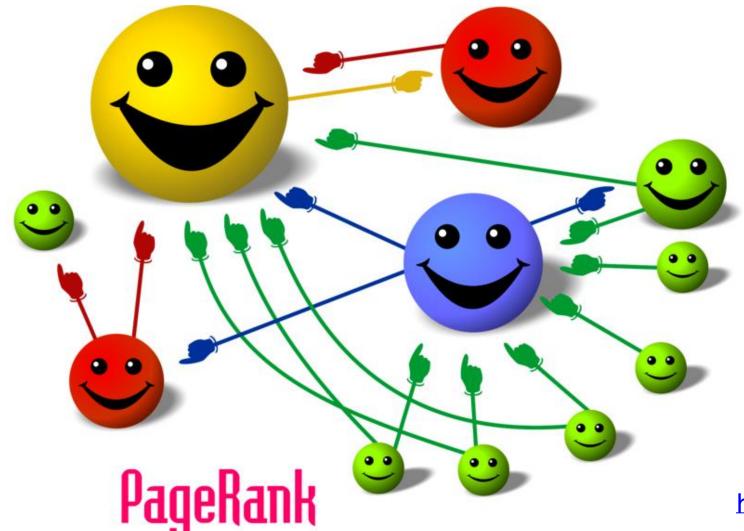
4 Directions and Weights

- 4.1 Directed Networks
- ▶ 4.2 The Web
- 4.3 PageRank
- 4.4 Weighted Networks
- 4.5 Information and Misinformation
- 4.6 Co-occurrence Networks
- 4.7 Weight Heterogeneity
- 4.8 Summary
- 4.9 Further Reading
- Exercises



4.3 PageRank / background

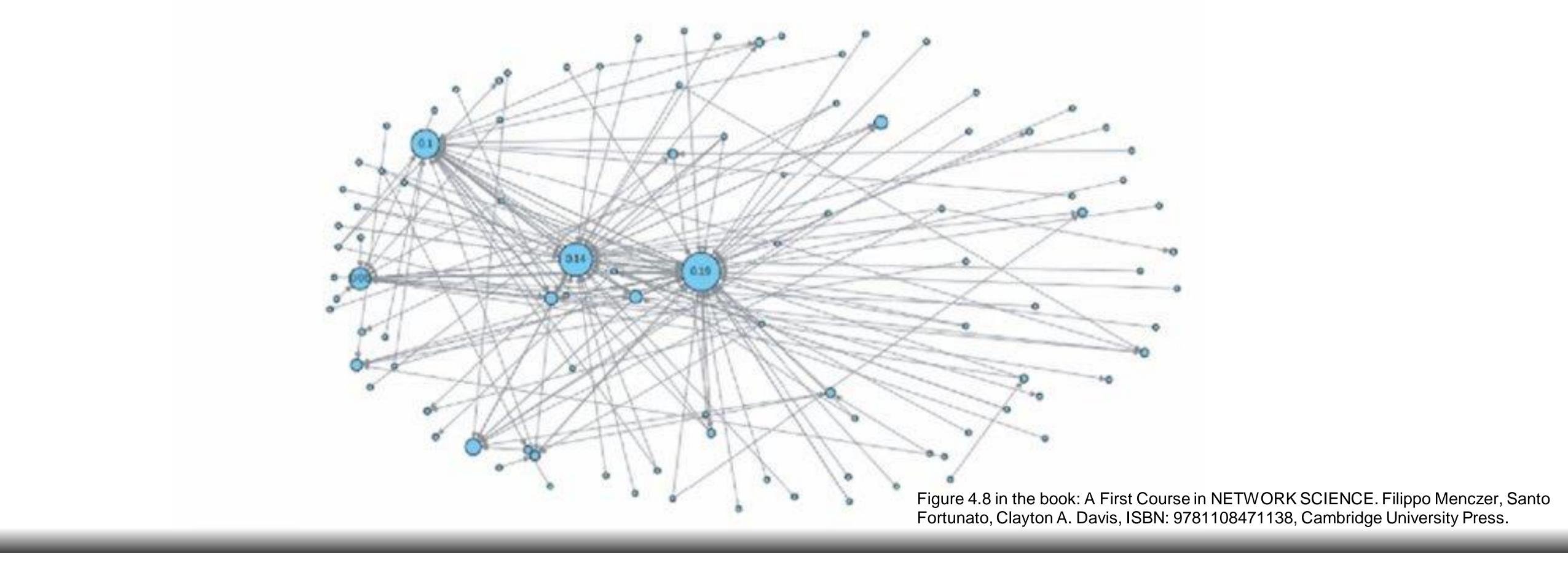
- When we submit a query, the search engine can quickly list all the matching pages based on a search index.
- Ranking algorithms are critical components of a search engine.
- Search engines use network centrality measures as ranking criteria.
- Example: PageRank introduced by Google in 1998.



https://en.wikipedia.org/wiki/File:PageRank-hi-res.png, CC-BY-SA

4.3 PageRank / definition

- PageRank is an algorithm to compute a centrality measure that aims to capture the prestige or importance of each node.
- It is typically used in directed networks.
- Example: a spammer's page created by copying from Wikipedia, and the Wikipedia page.



The size of the nodes is proportional to their PageRank.



4.3 PageRank / searching behaviors

- Random walk or random surf: random link from page to page
- Teleportation or random jump: user starts a new browsing session.

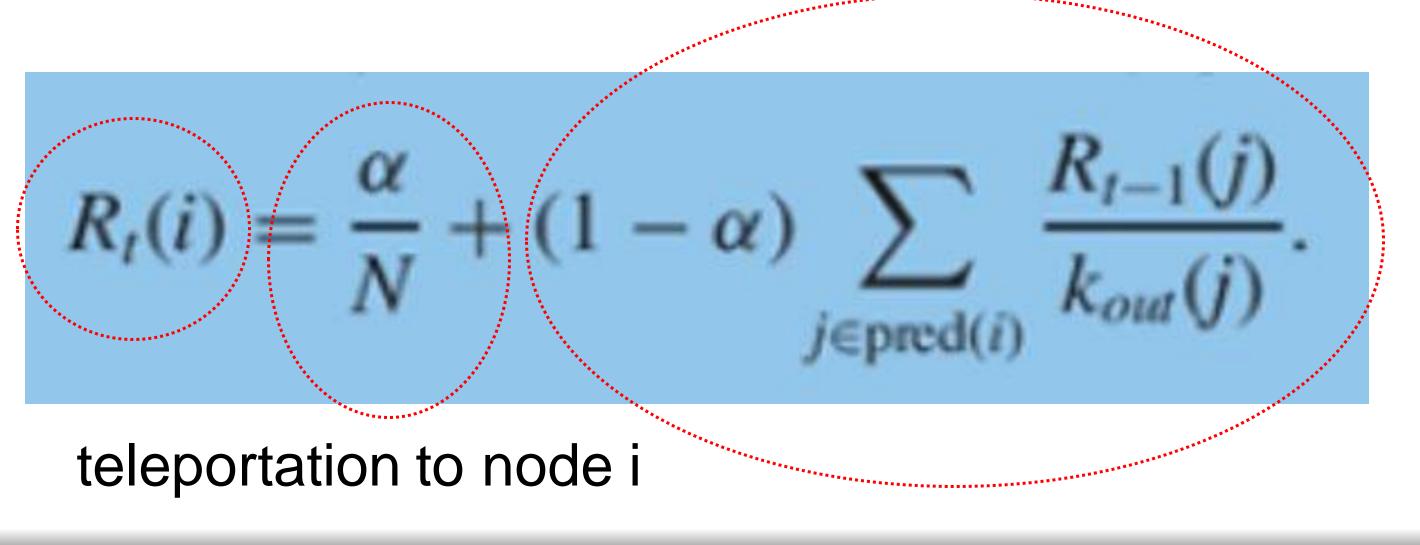
$$R_t(i) = \frac{\alpha}{N} + (1 - \alpha) \sum_{j \in \text{pred}(i)} \frac{R_{t-1}(j)}{k_{out}(j)}.$$

E.q. 4.1 in the book: A First Course in NETWORK SCIENCE. Filippo Menczer, Santo Fortunato, Clayton A. Davis, ISBN: 9781108471138, Cambridge University Press.

4.3 PageRank / how to compute

It is computed with an interactive approach called the power method.

PageRank of node i at time t



How one can traverse one of the links to i during random walk

4.3 PageRank / how to compute

$$R_t(i) = \frac{\alpha}{N} + (1 - \alpha) \sum_{j \in \text{pred}(i)} \frac{R_{t-1}(j)}{k_{out}(j)}.$$

4.3 PageRank / how to compute

- Interactive demonstration of PageRank using NetLogo (Appendix B.1)
- The pagerank() function from NetworkX package:

PR = nx.pagerank(D) # D is a DiGraph

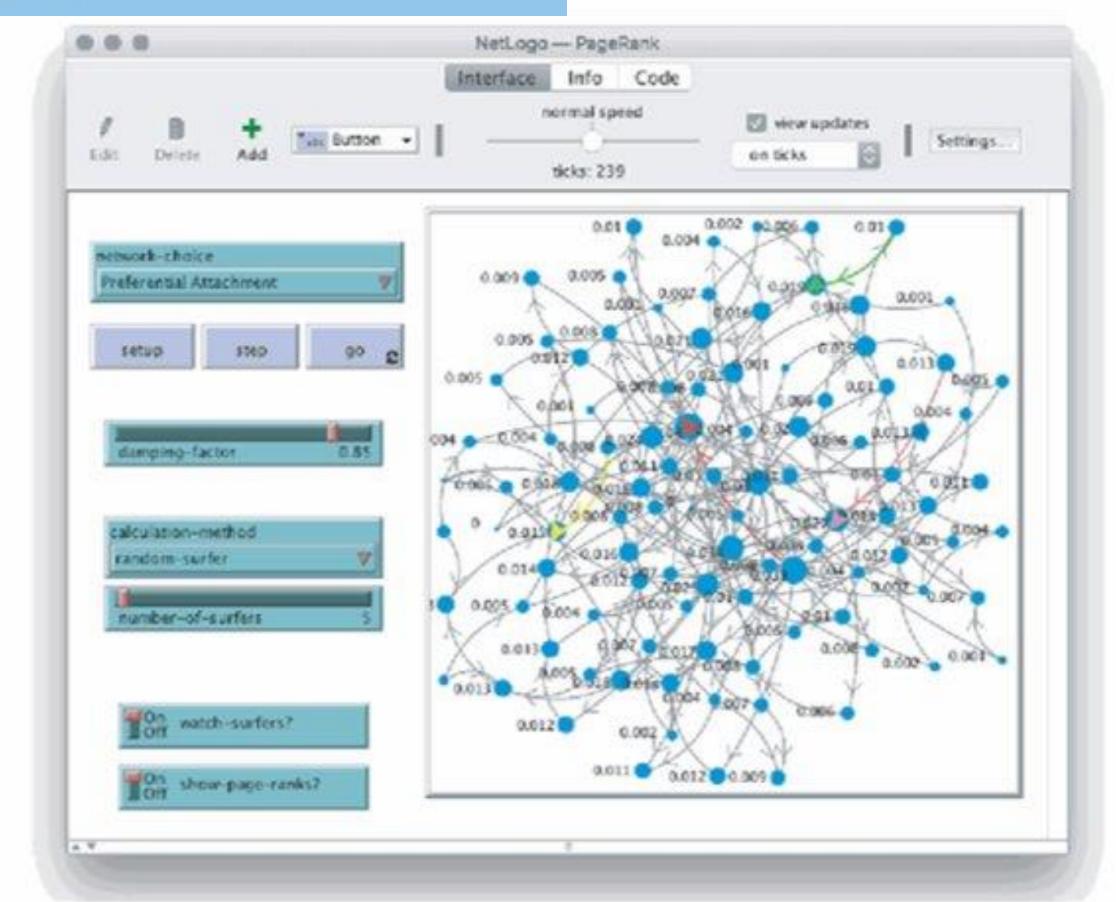
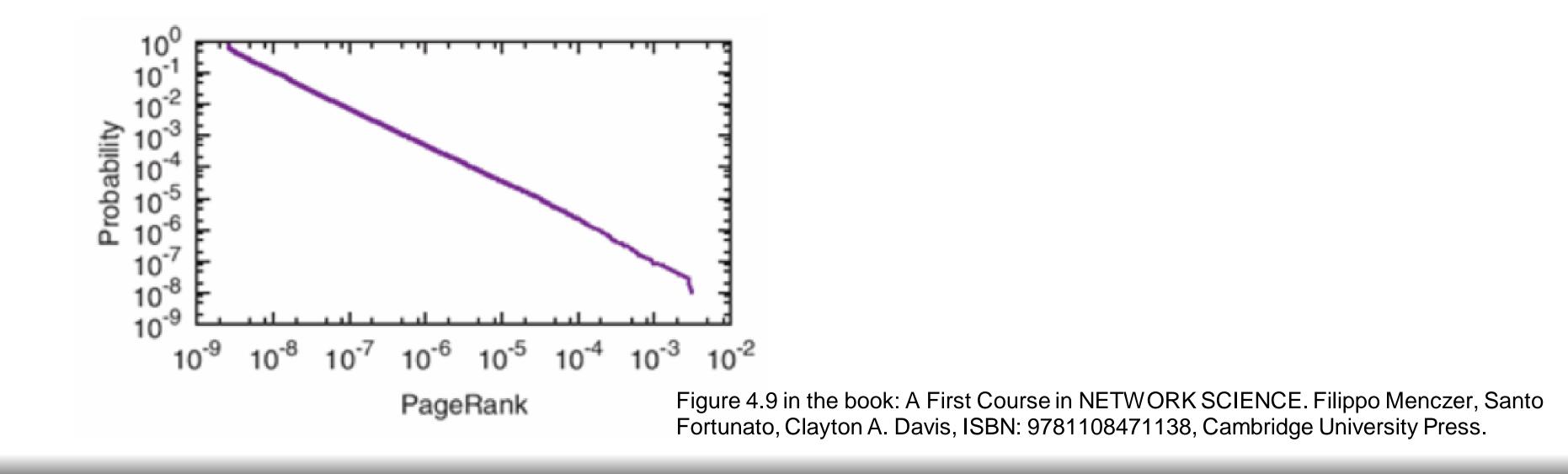


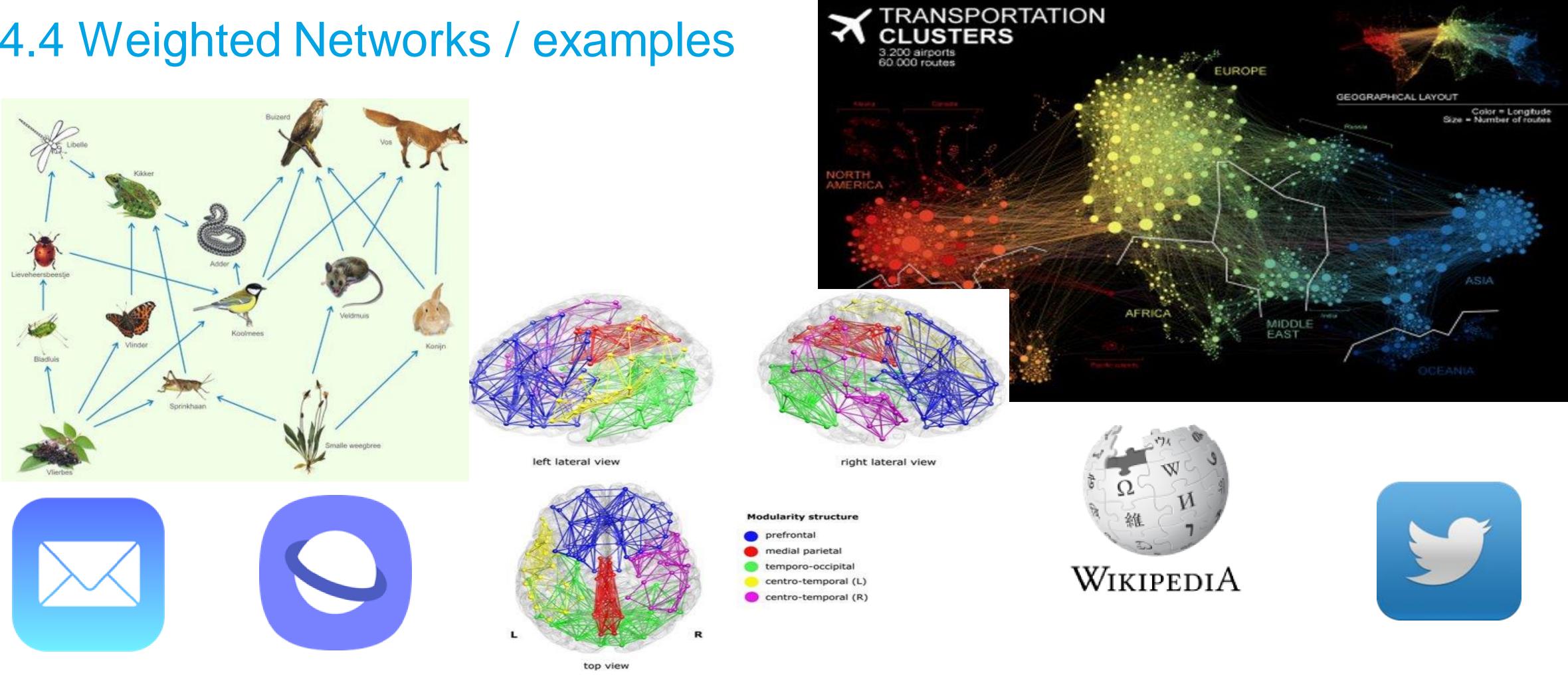
Figure B.1 in the book: A First Course in NETWORK SCIENCE. Filippo Menczer, Santo Fortunato, Clayton A. Davis, ISBN: 9781108471138, Cambridge University Press.



4.3 PageRank vs in-degree

- The distribution of PageRank is similar to the distribution of in-degree on the Web.
- Among two pages with the same in-degree, the one linked by pages with higher PageRank wins the game.
- Search Engine Optimization (SEO) helps websites improve their search ranking.

4.4 Weighted Networks / examples

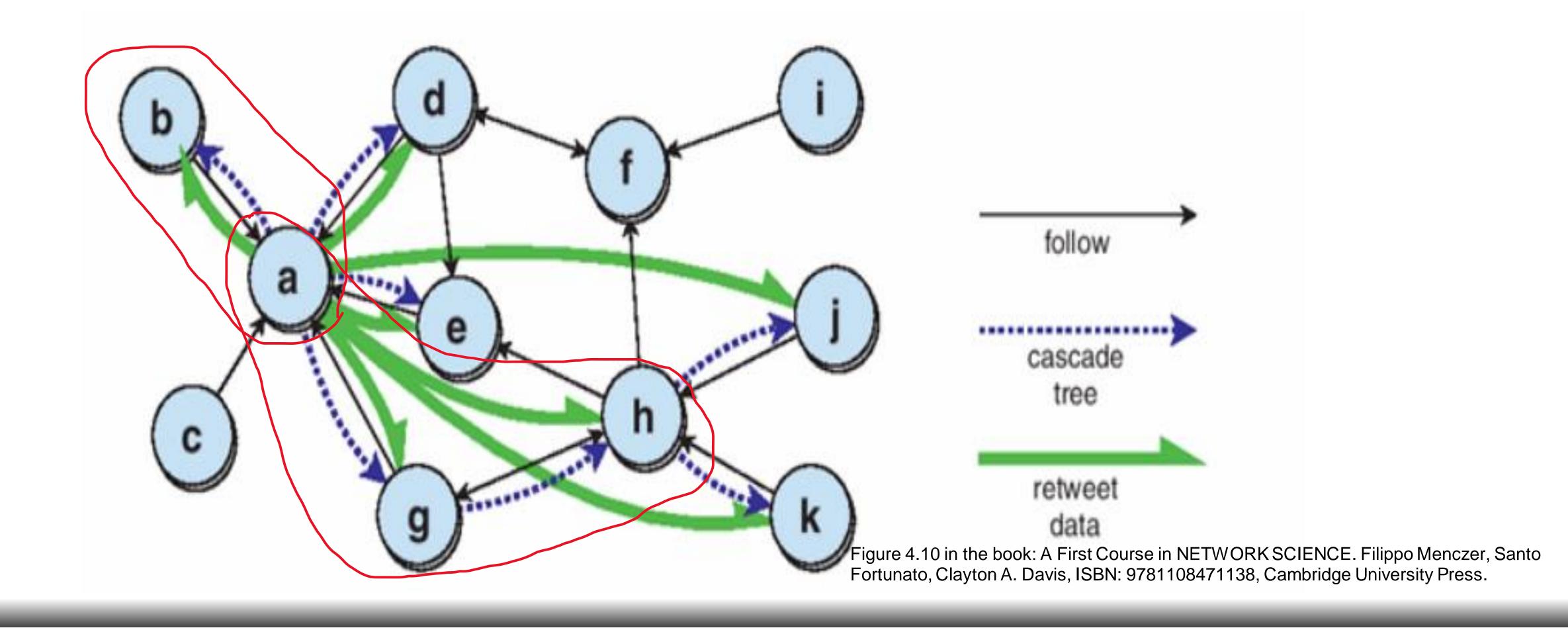


Node centrality measures are strength, in-strength, out-strength.

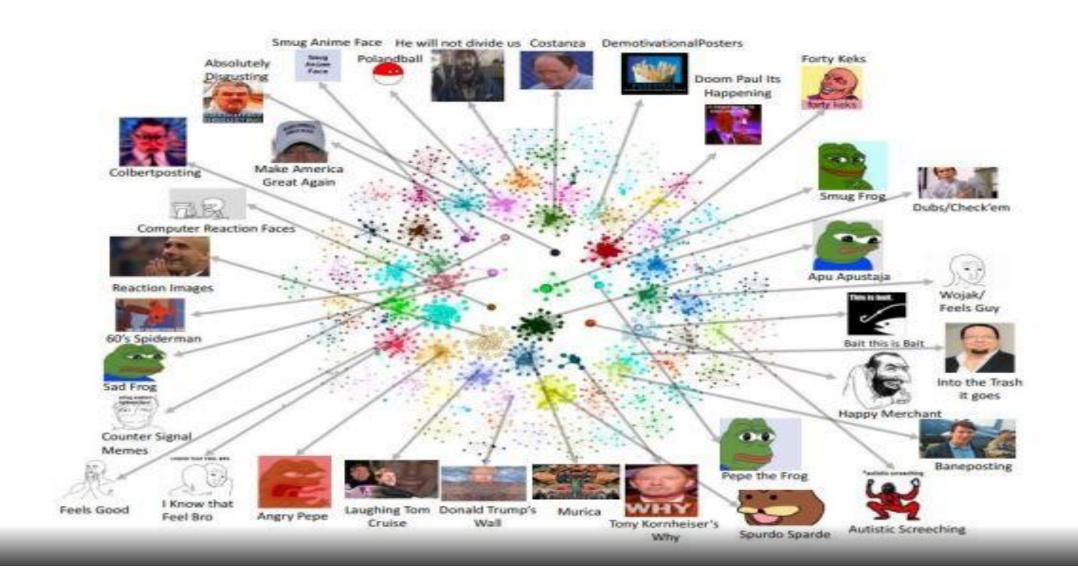


4.5 Information and Misinformation / diffusion networks

- Diffusion network is a case study to explore the features of weighted directed networks.
- Related terms: trees, cascade trees (star trees), forest (set of trees), multilayer networks, and temporal networks (section 2 in the book).

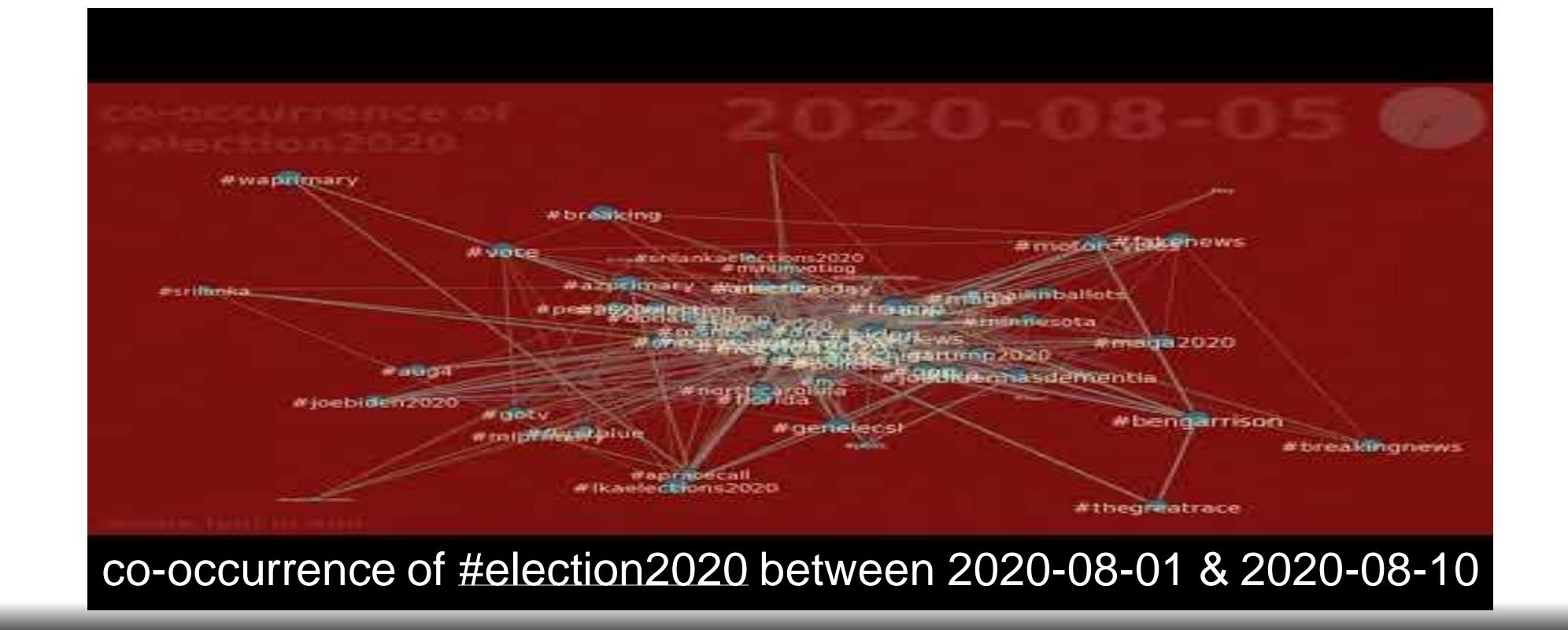


Follower and retweet networks on Twitter.



4.5 Information and Misinformation / information diffusion networks

- People are nodes and pieces of information that are passed from a person to person are links.
- Example, Twitter diffusion networks are obtained by aggregating cascade forests across time and across many popular hashtags.



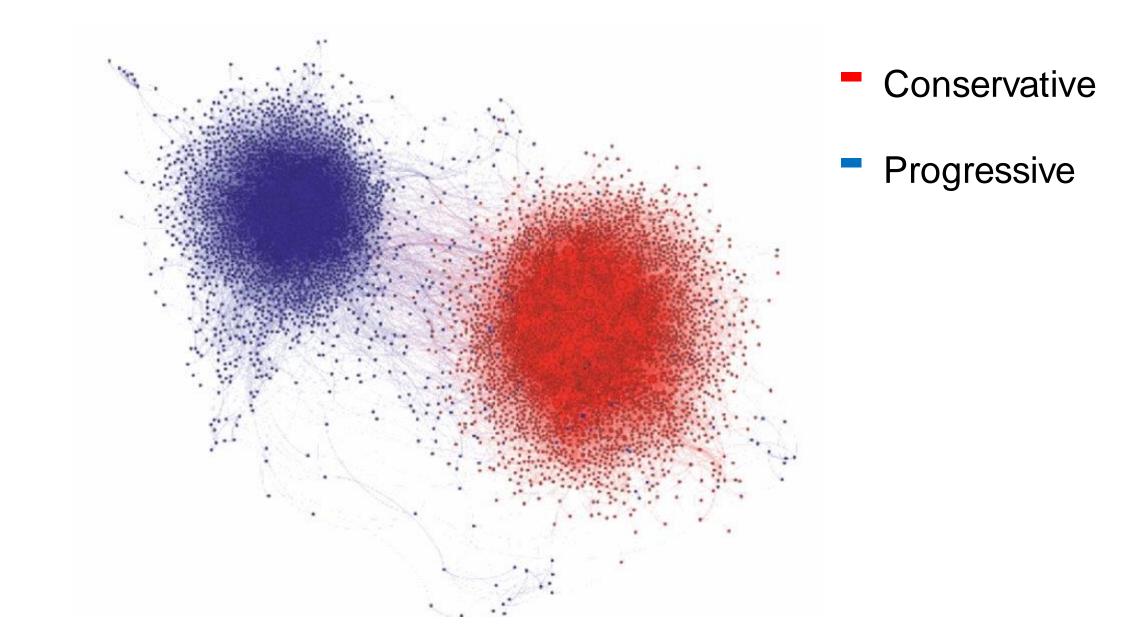
4.5 Information and Misinformation / diffusion networks / examples

Explore interactive diffusion networks from Twitter at

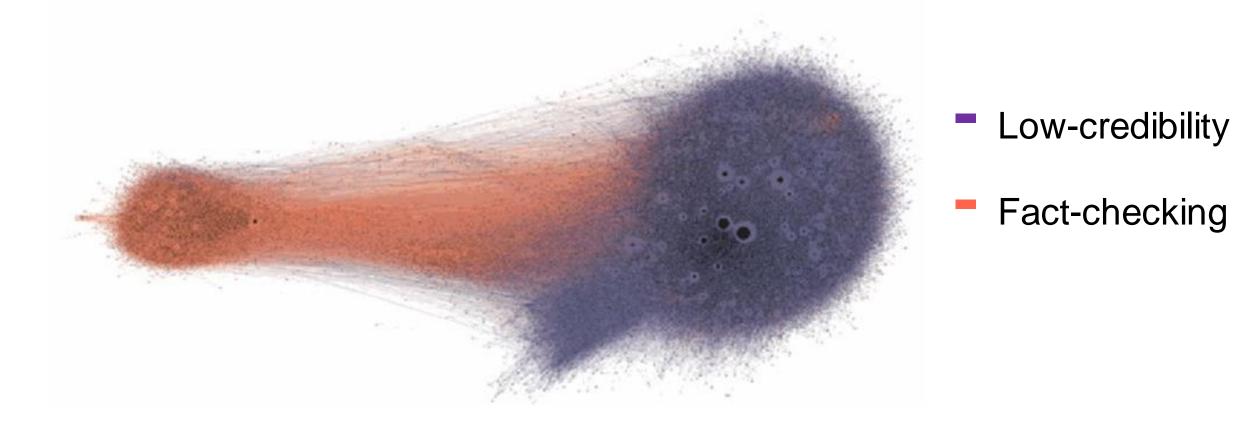
http://osome.iuni.iu.edu/tools/movies/ (INDIANA UNIVERSITY OBSERVATORY ON SOCIAL MEDIA)

4.5 Information and Misinformation / features of diffusion networks

Echo chambers: is the spread uniform across the network or concentrated with dense clusters of network?



A user is mostly exposed to opinions reinforcing their own.

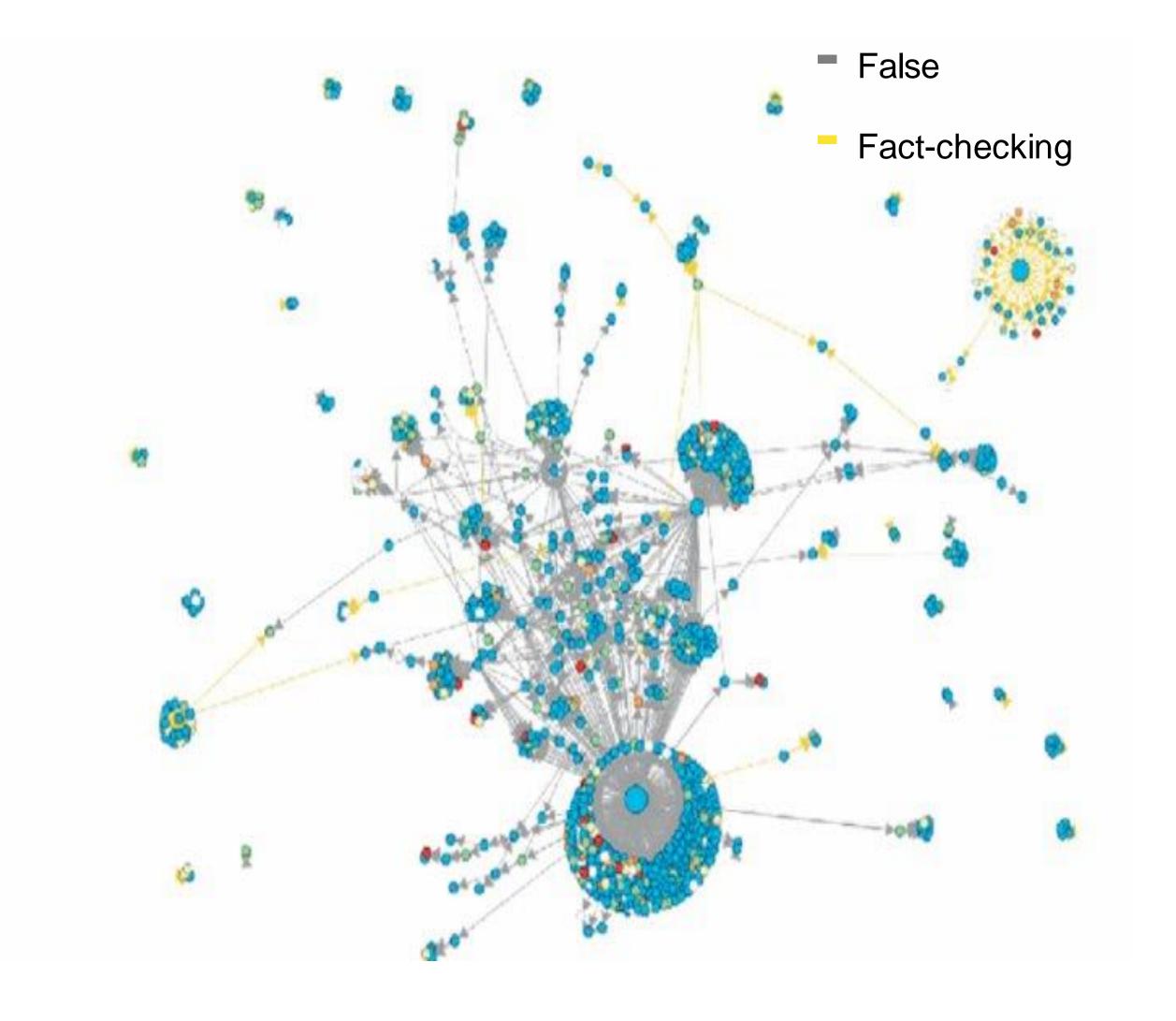


People who are vulnerable to political misinformation.

Figures 0.3 and 4.11 in the book: A First Course in NETWORK SCIENCE. Filippo Menczer, Santo Fortunato, Clayton A. Davis, ISBN: 9781108471138, Cambridge University Press.

4.5 Information and Misinformation / features of diffusion networks

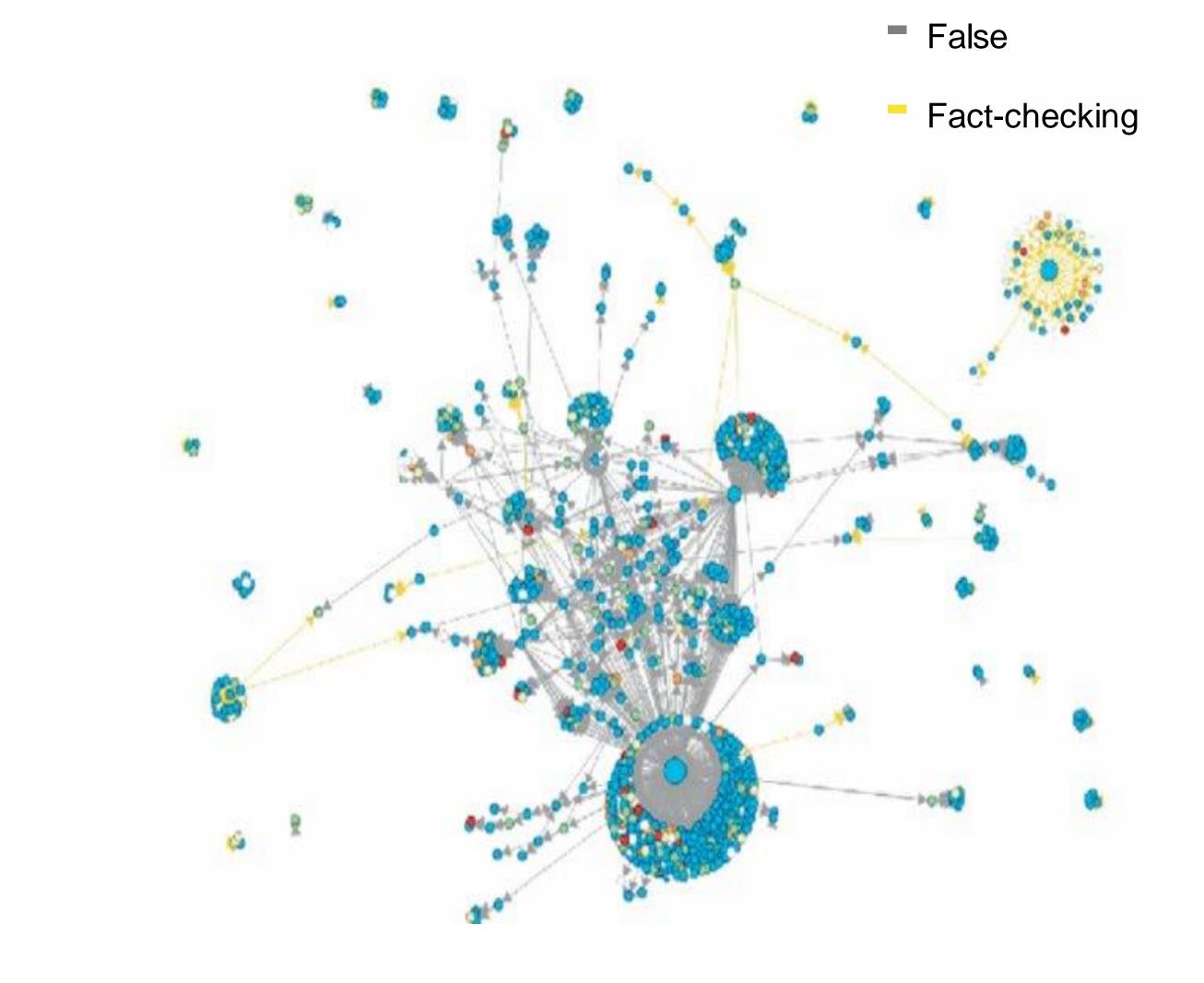
Virality: It can be illustrated by the number of users exposed to a message.



The spread of misinformation is more viral than actual news.

4.5 Information and Misinformation / features of diffusion networks

Influence: a high value of out-strength.

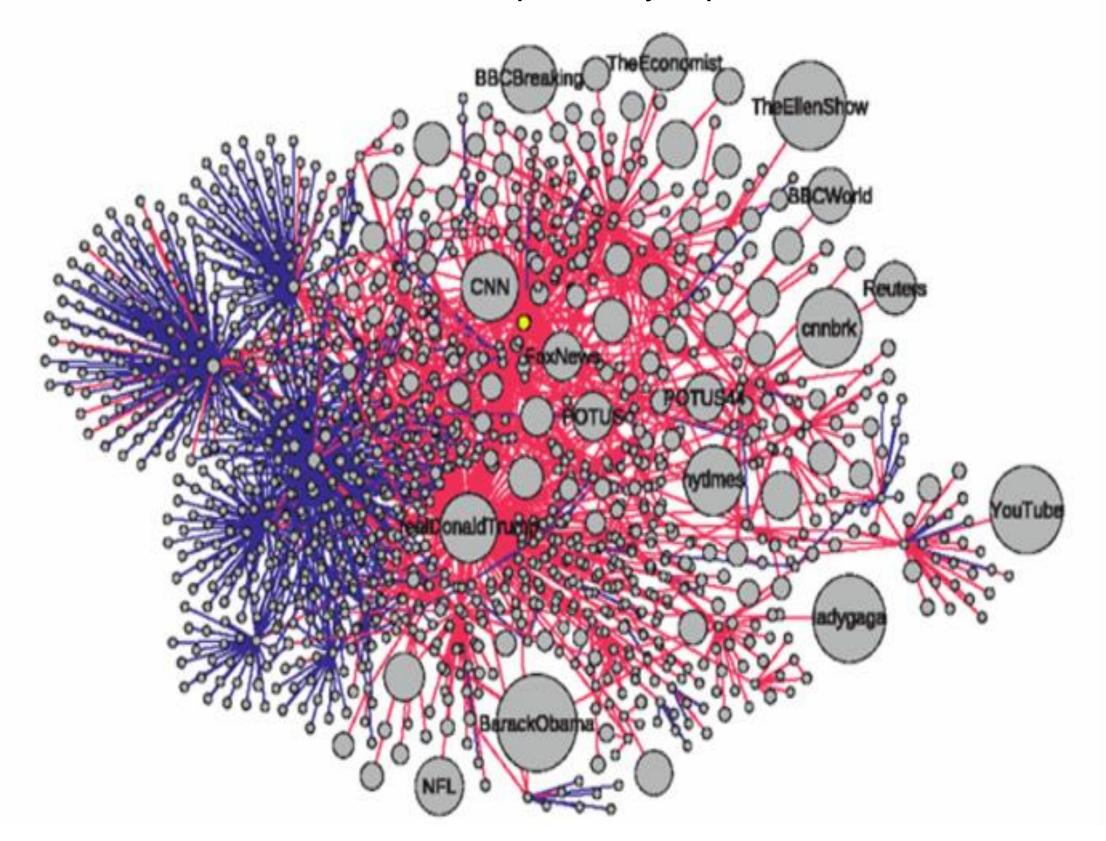


Node size is proportional to out-strength.

4.5 Information and Misinformation / social bots

Bots can gain significant influence.

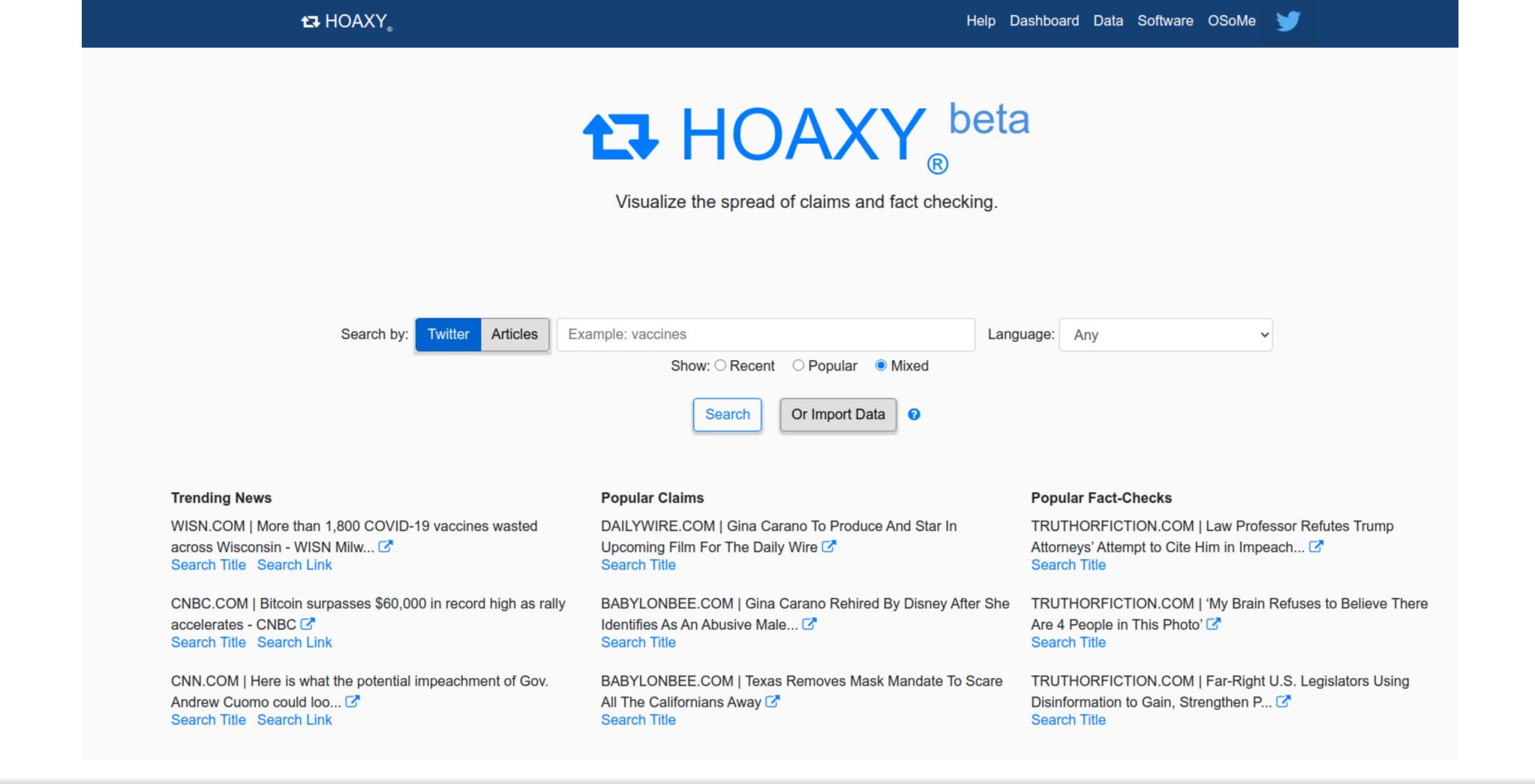
- links, the article spread by retweet or quoted tweet
- links, the article spread by replies or mentions.



Node size represents the number of followers of an account. Yellow node is a bot.



Explore Misinformation tools at http://osome.iuni.iu.edu/tools/ (INDIANA UNIVERSITY OBSERVATORY ON SOCIAL MEDIA)



Explore the role of bots https://hoaxy.osome.iu.edu/.

Summary:

Search engines use network centrality measures as ranking criteria.

 In weighted networks, node centrality measures are strength, instrength, out-strength.

 Features of diffusion networks such as echo chambers, virality, and influence are used in investigating online misinformation.

Exercise: Chapter 4 tutorial notebook.

https://github.com/CambridgeUniversityPress/FirstCourseNetworkScience/blob/master/tutorials/Chapter%204%20Tutorial.ipynb