

Socio-Semantic Networks and the Structural Space

Adina Nerghes

Digital Humanities Lab

- Advancing the humanities through digital methods
- Focus on big ‘textual’ data
- Interdisciplinary
 - Events & Entities
 - Change
 - Connections



Marieke van Erp



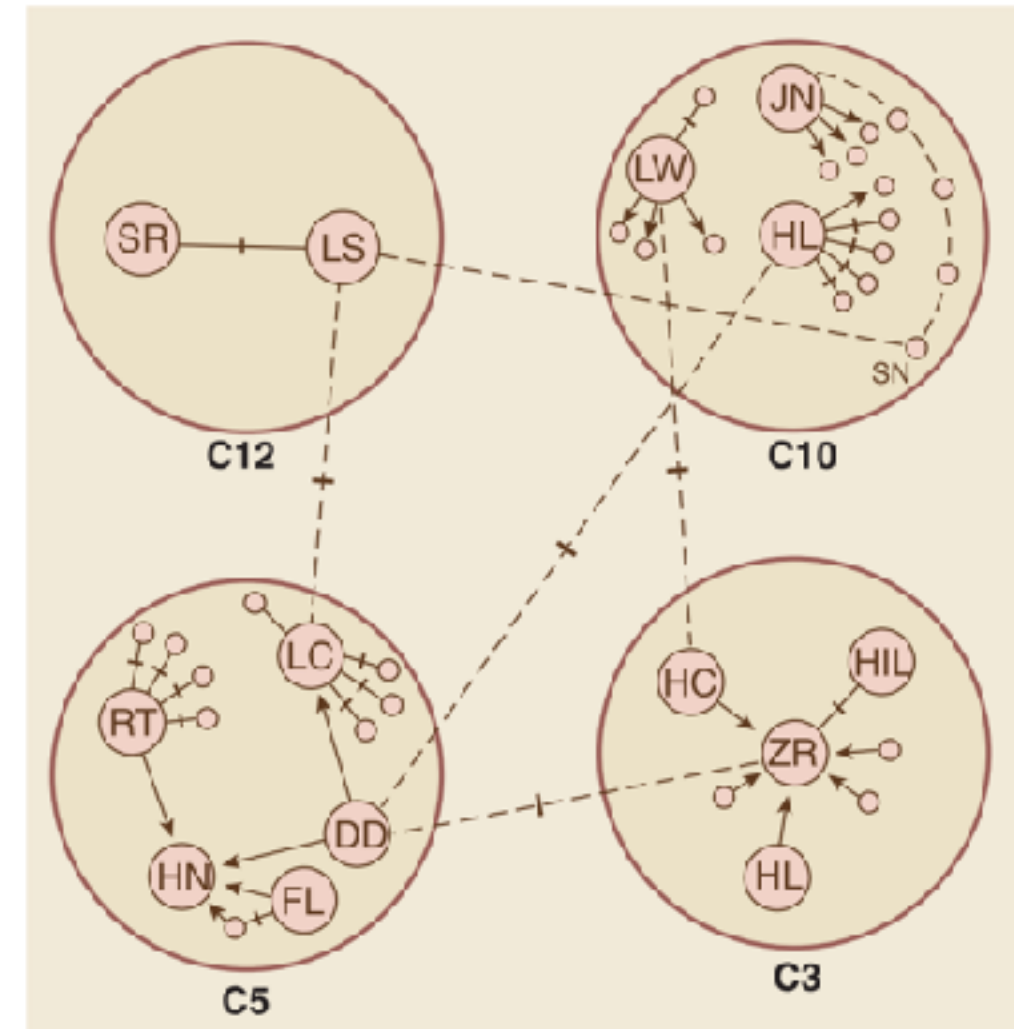
Melvin Wevers



Adina Nerghes

Jacob Moreno and the Birth of Social Network Analysis

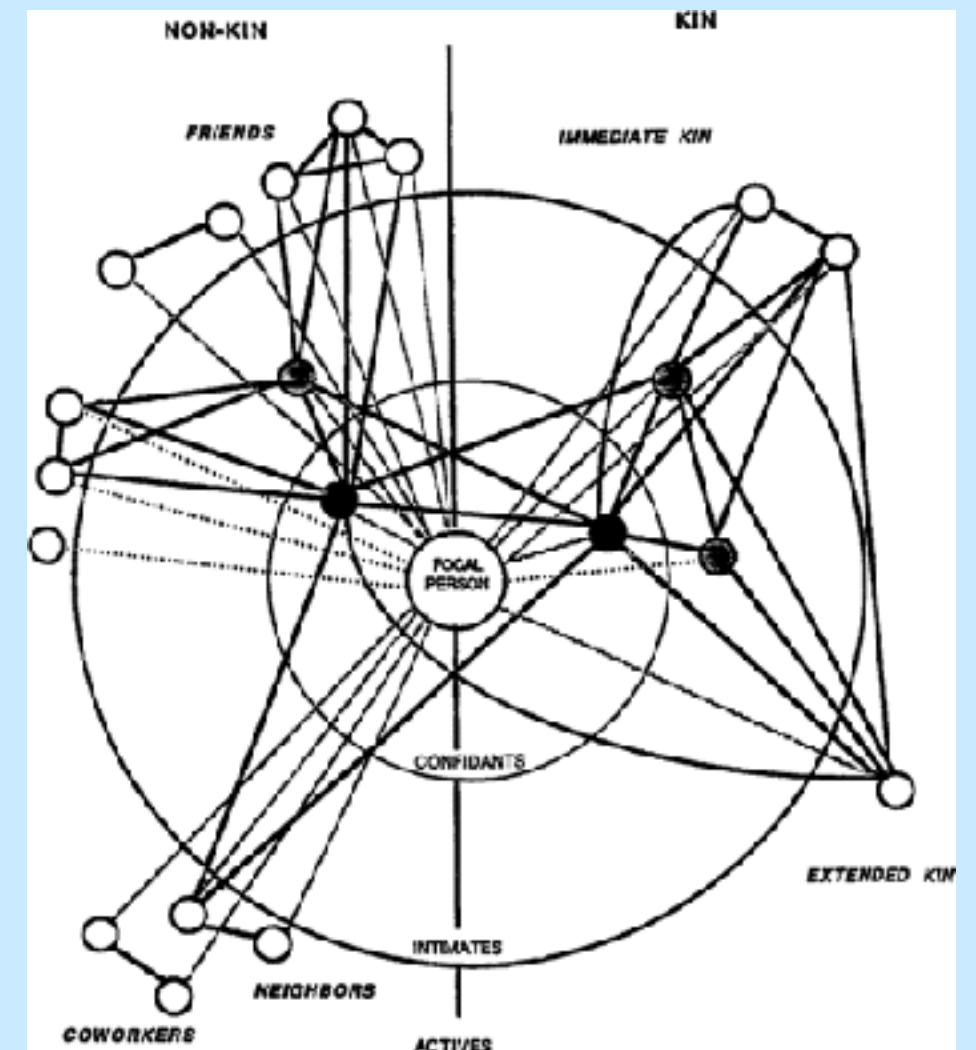
- Psychiatrist and psychodrama founder
- Moreno conducted studies to research group behavior using "sociometric tests"
- The four features of social network analysis (Freeman, 2004):
 - Motivated by a structural intuition based on ties linking social actors
 - Grounded in systematic empirical data
 - Use of graphic imagery
 - Relied on the use of mathematical and/or computational models



Moreno's network of runaways.

A relational and structural perspective

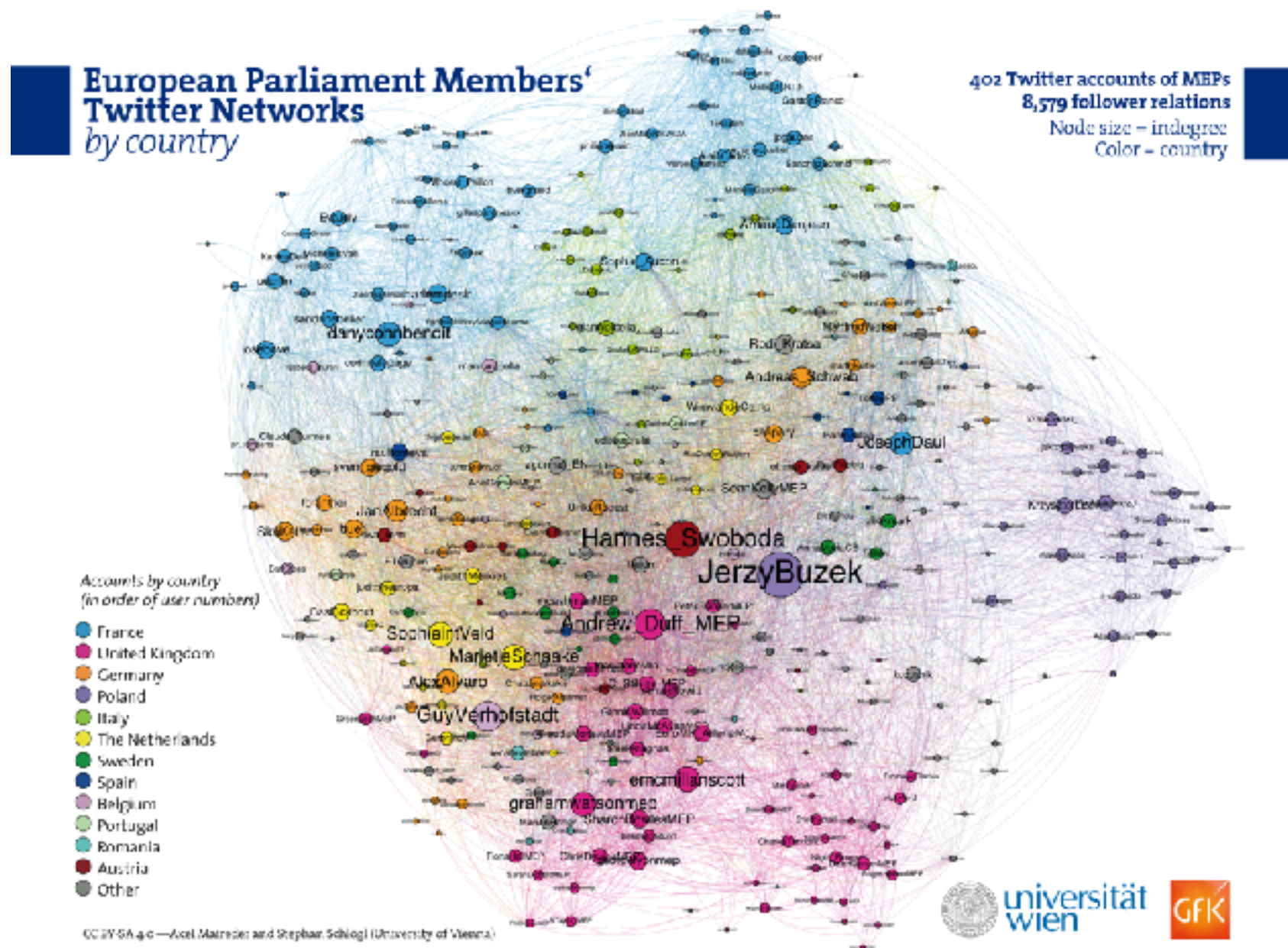
- The network perspective involves the study of entities as embedded in a network of **relations** and seek explanations for social behavior in the **structure** of these networks rather than in the individuals alone
- **Not just a methodology:** it is a unique perspective on how society functions. Instead of focusing on individuals and their attributes, or on macroscopic social structures, it centers on **relations**
- Applications:
 - Understand how to improve the **effectiveness** of a network
 - Uncover **patterns** in relationships or interactions
 - Find/follow **paths** that information flows
 - Identify **key** players
 - Test **hypotheses**
 - Promote social **cohesion**



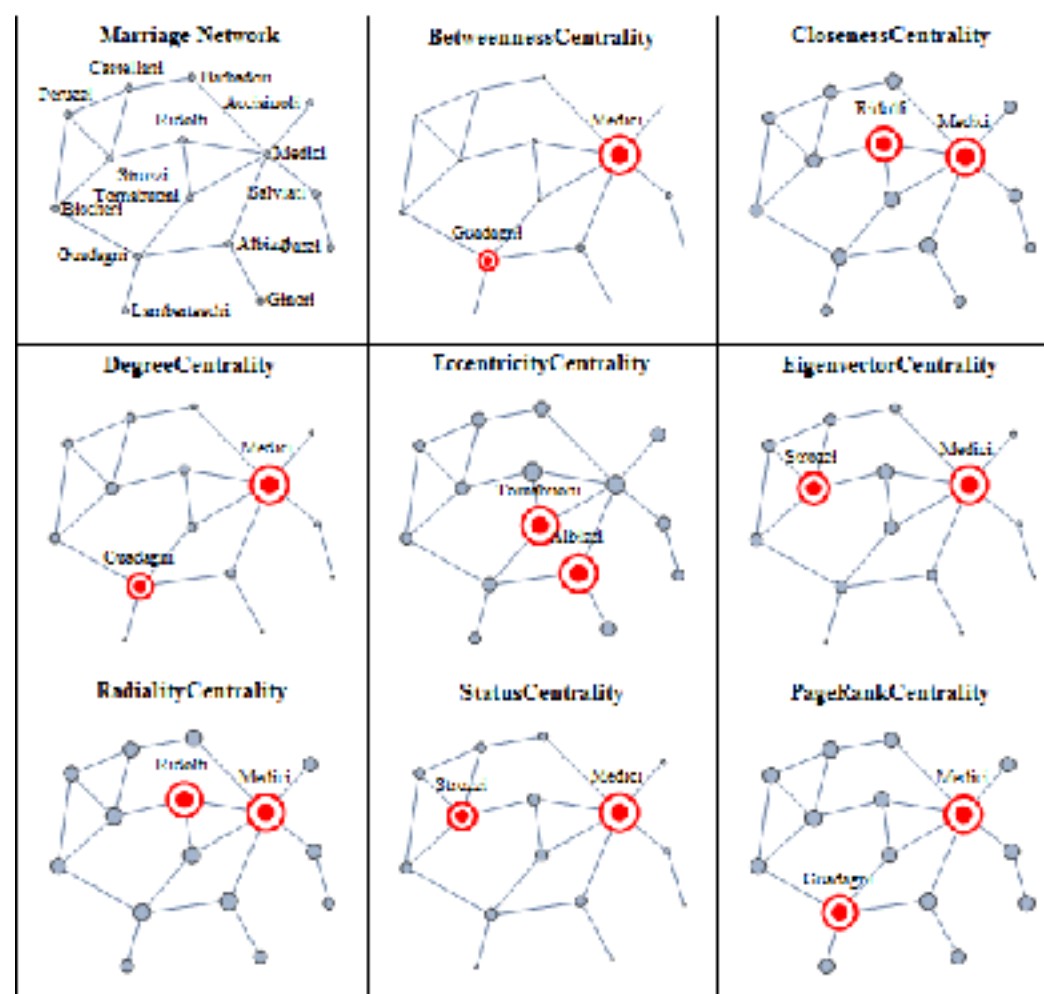
This is an early depiction of what we call an 'ego' network, i.e. a personal network. The graphic depicts varying tie strengths via concentric circles (Wellman, 1998)

Similarities			Social Relations				Interactions	Flows
Location e.g., Same spatial and temporal space	Membership e.g., Same clubs Same events etc.	Attribute e.g., Same gender Same attitude etc.	Kinship e.g., Mother of Sibling of	Other role e.g., Friend of Boss of Student of Competitor of	Affective e.g., Likes Hates etc.	Cognitive e.g., Knows Knows about Sees as happy etc.	e.g., Sex with Talked to Advice to Helped Harmed etc.	e.g., Information Beliefs Personnel Resources etc.

(Borgatti et. al., 2009)

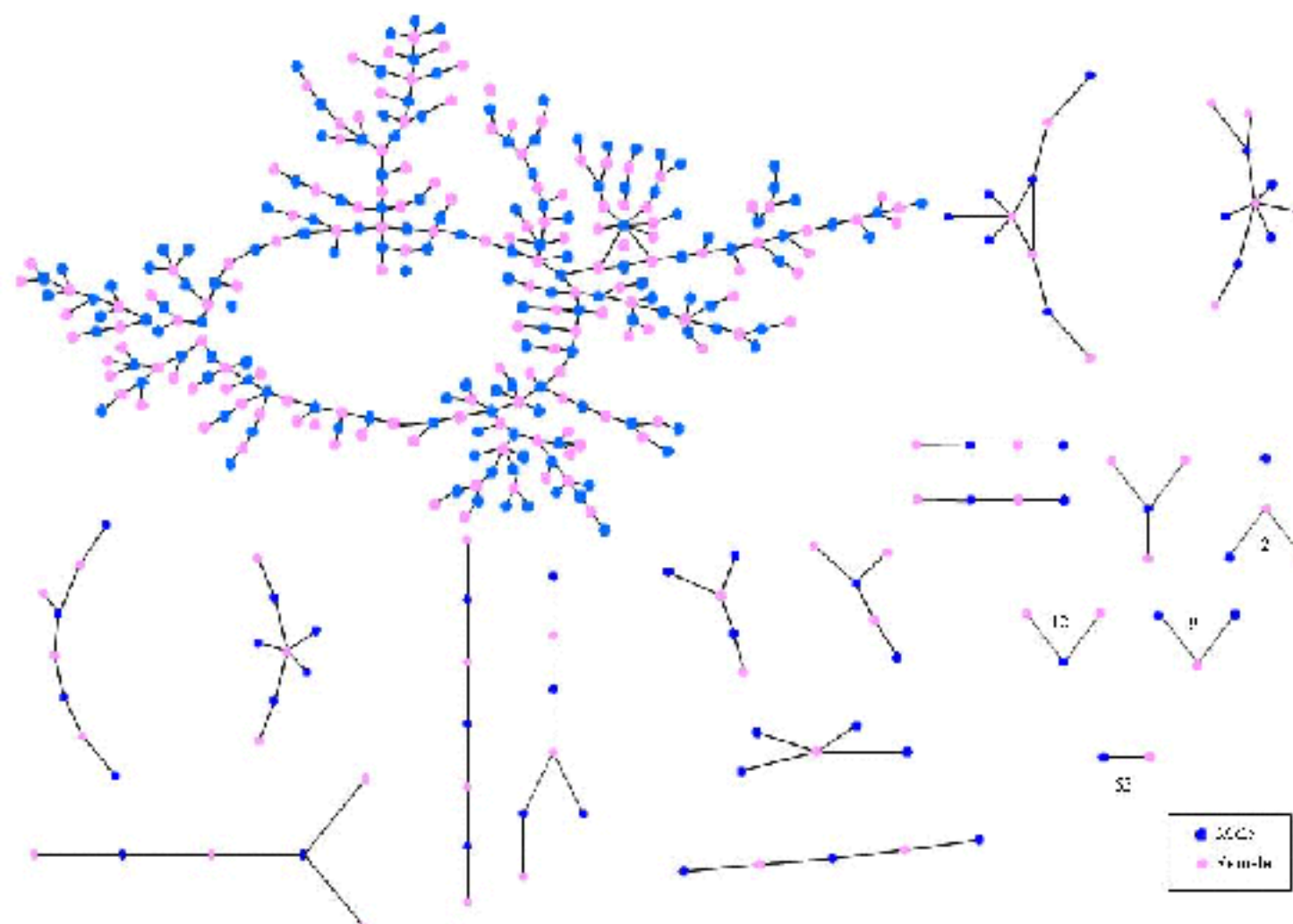


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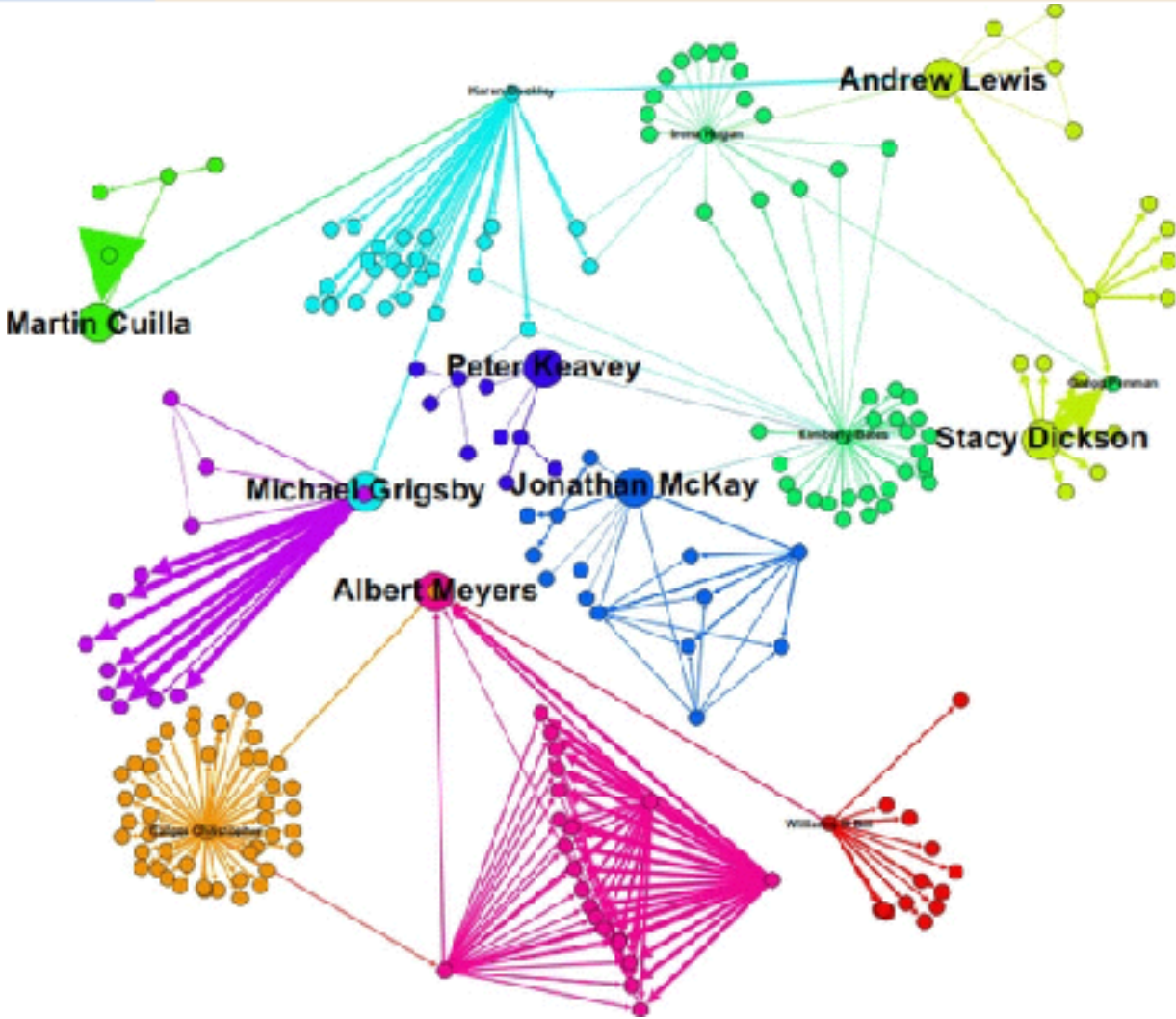


Centrality and Prestige of Florentine Families

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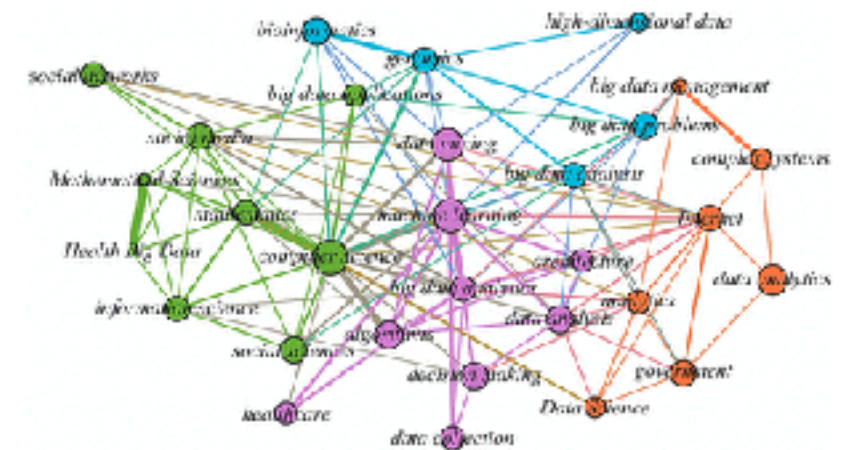
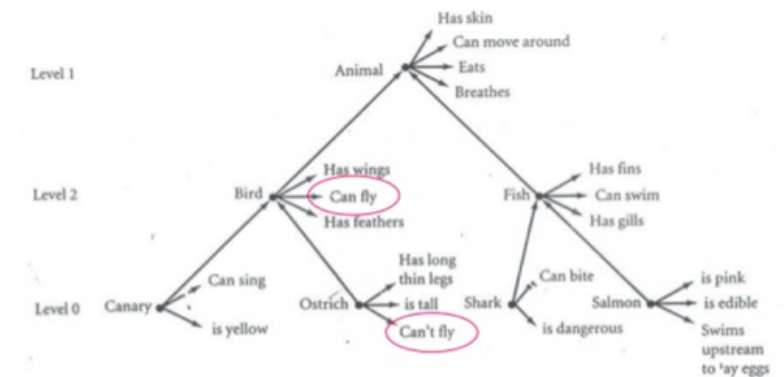
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A reduced social network depicting email flow from group leaders in a number of groups in Enron.

Semantic networks

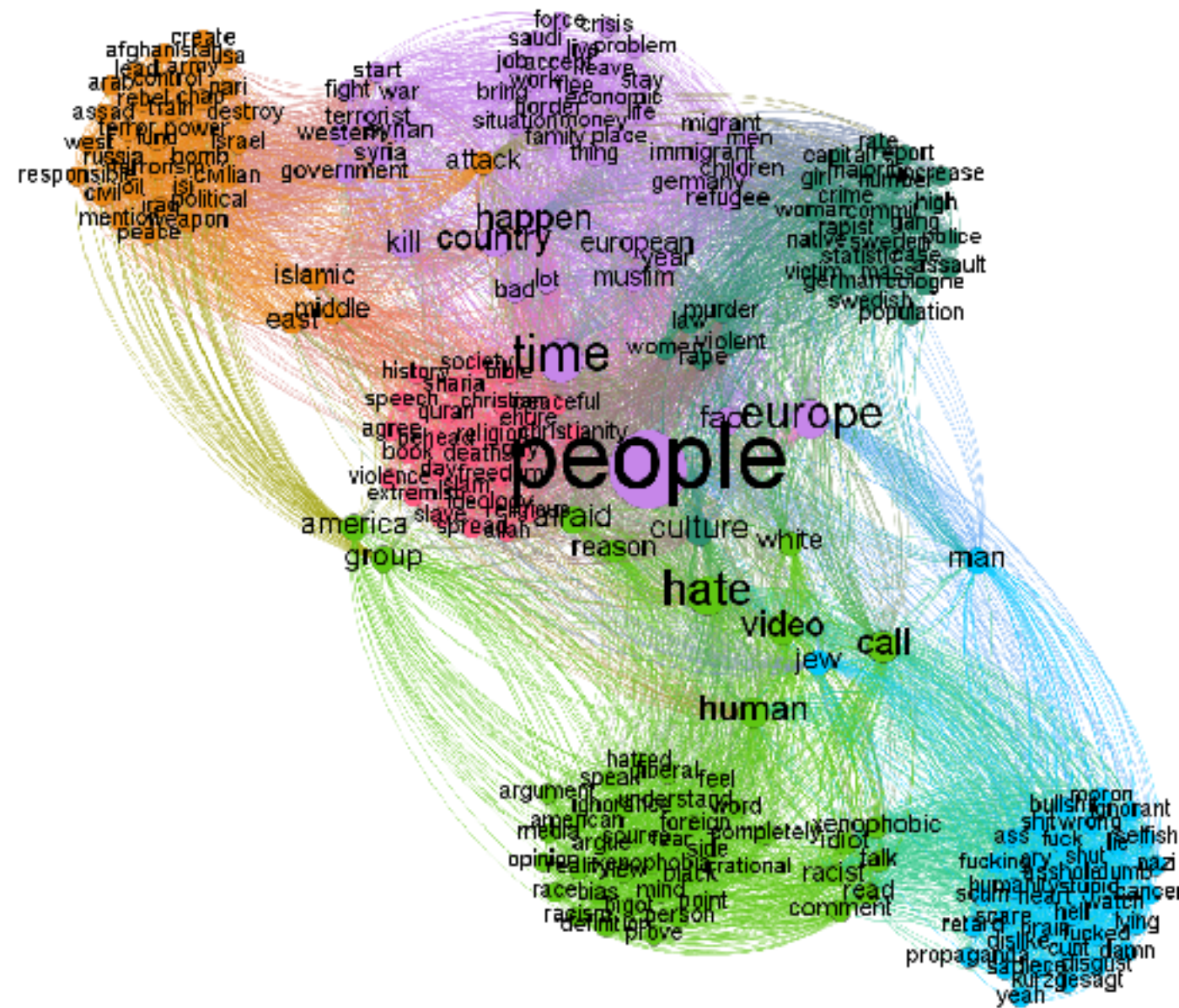
- Defined as: "representational format [that would] permit the 'meanings' of words to be stored, so that humanlike use of these meanings is possible" (Quillian, 1968, p. 216)
- The meaning of a word could be represented by the set of its verbal associations
- Basic assumption: language (is) can be modeled as networks of words and the (lack of) relations among words



Networks of words

Semantic Networks

Networks of concepts



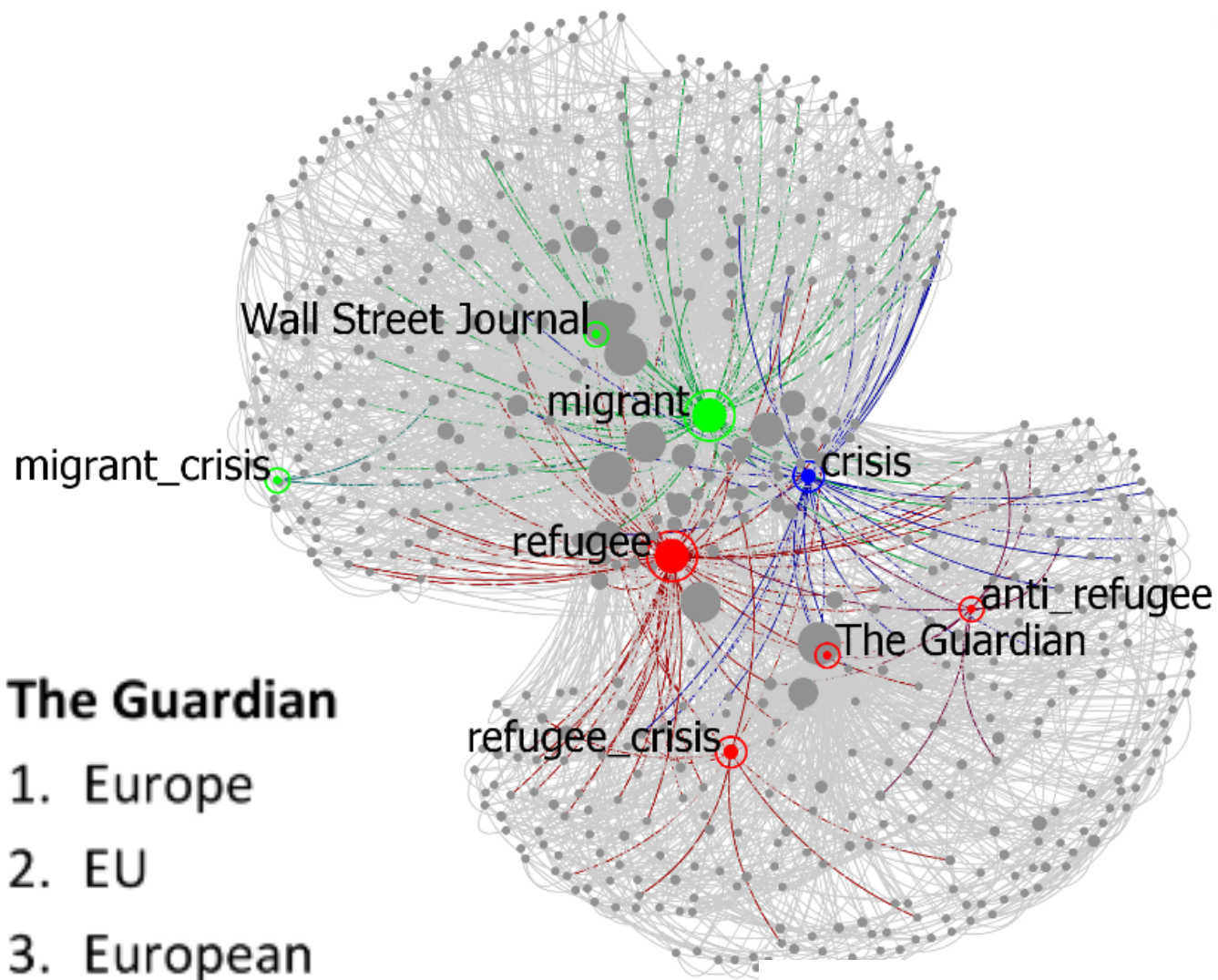
Content networks

Co-word maps

Maps

What makes semantic networks interesting?

- Correspond to a natural way of **organizing information** and the way humans think
- Semantic networks allow to **model** semantic relationships (Sowa, 1991)
- Investigate the meaning of texts by detecting the **relationships** between and among words and themes (Alexa, 1997; Carley, 1997a)
- Allow the analysis of words in their **context** (Honkela, Pulkki, & Kohonen, 1995)
- Expose semantic **structures** in document collections (Chen, Schuffels, & Orwig, 1996)
- Very **flexible** way of organizing data: you can easily **extend** the structure of semantic networks if needed
- You can easily convert **almost any** other data structure into semantic networks
- To represent **knowledge** or to support automated systems for reasoning about knowledge.

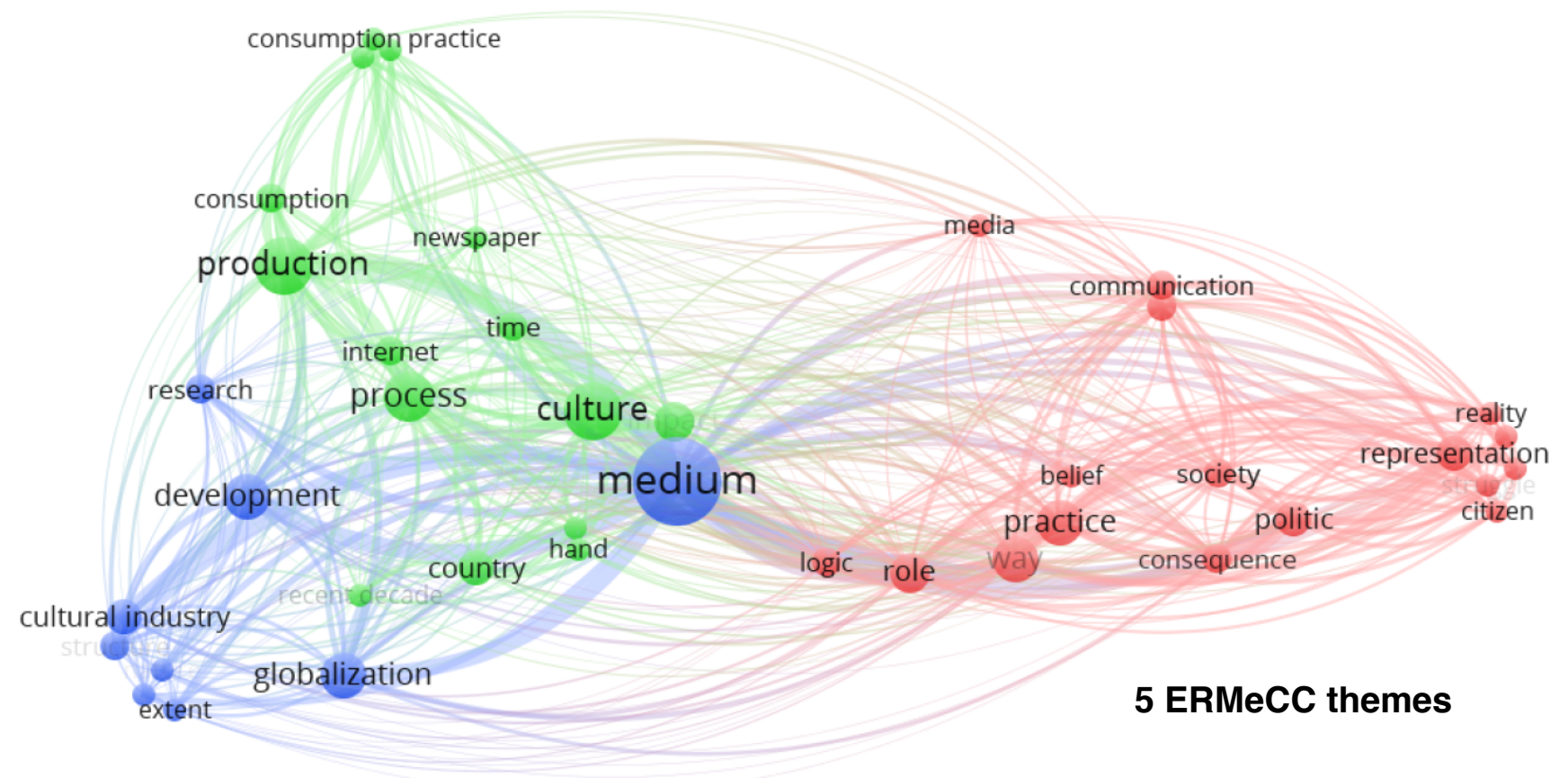


Wall Street Journal

1. Balkan
2. People
3. Asylum
4. Germany
5. Asylum seeker

The Guardian

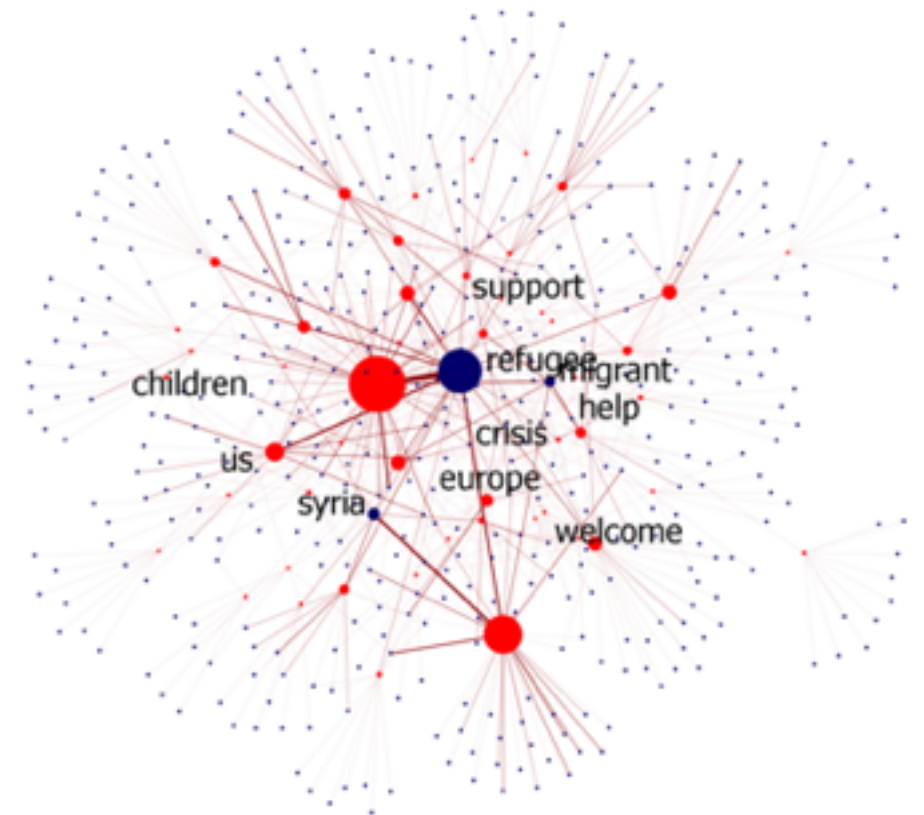
1. Europe
2. EU
3. European
4. Poland
5. Bad



5 ERMeCC themes

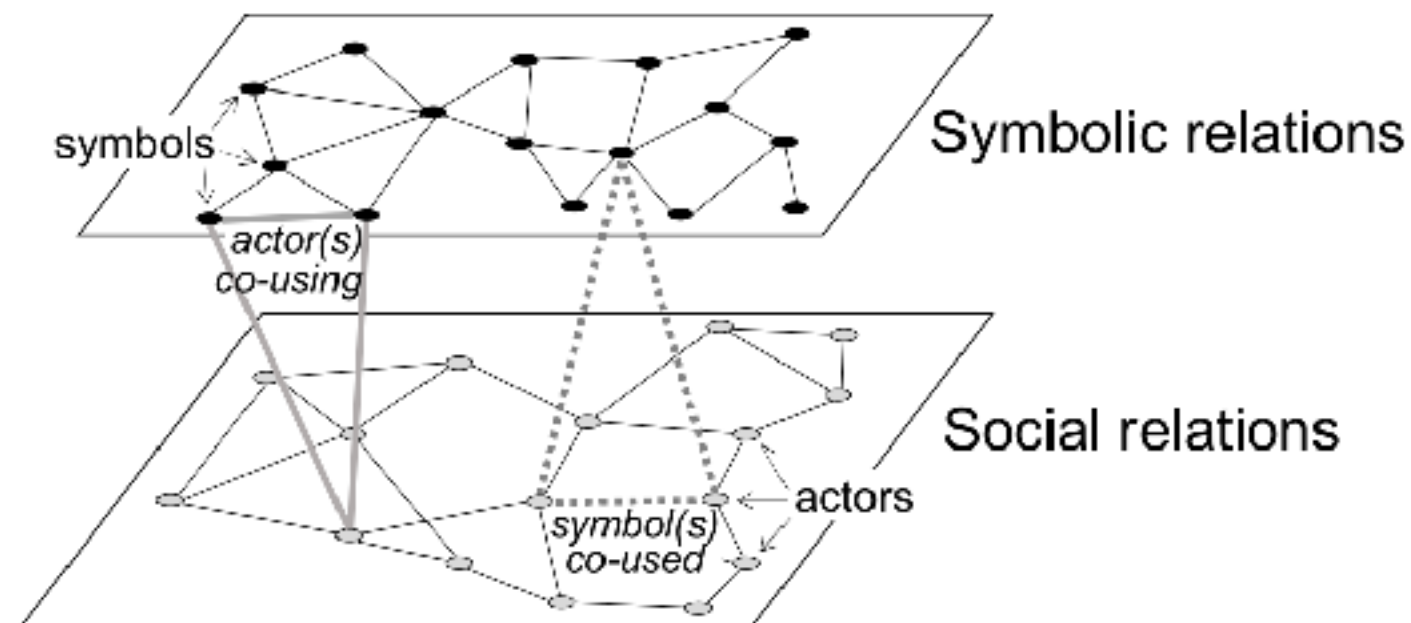
Socio-semantic networks

- The socio-semantic framework can account for the **meaning** structure along with the underlying **social** structures
- Can map not only how **meaning** is created through word co-use but also map this to the **pattern** of users connected to words and user interactions
- Based on 2-mode networks:
 - two-mode representations of actors/entities and the concepts they employ



Socio-semantic networks

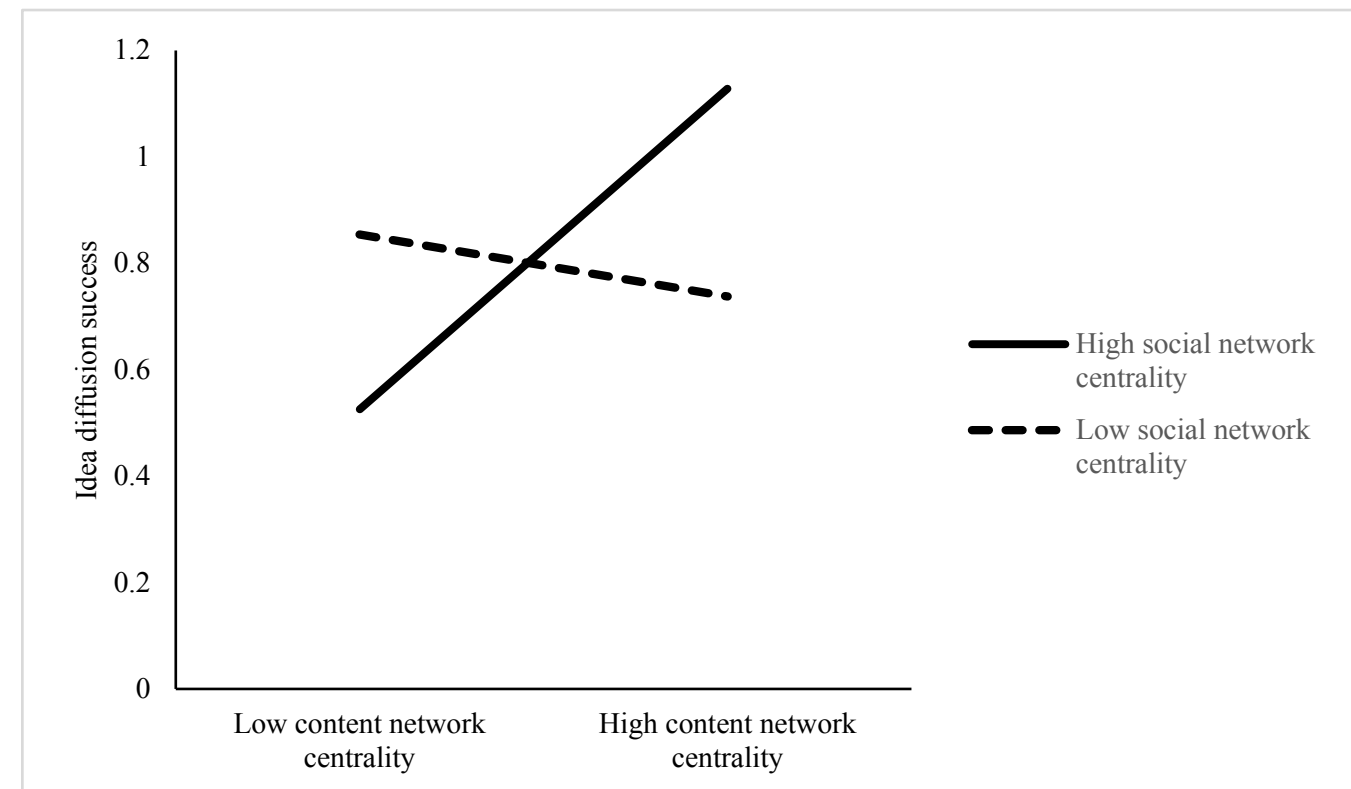
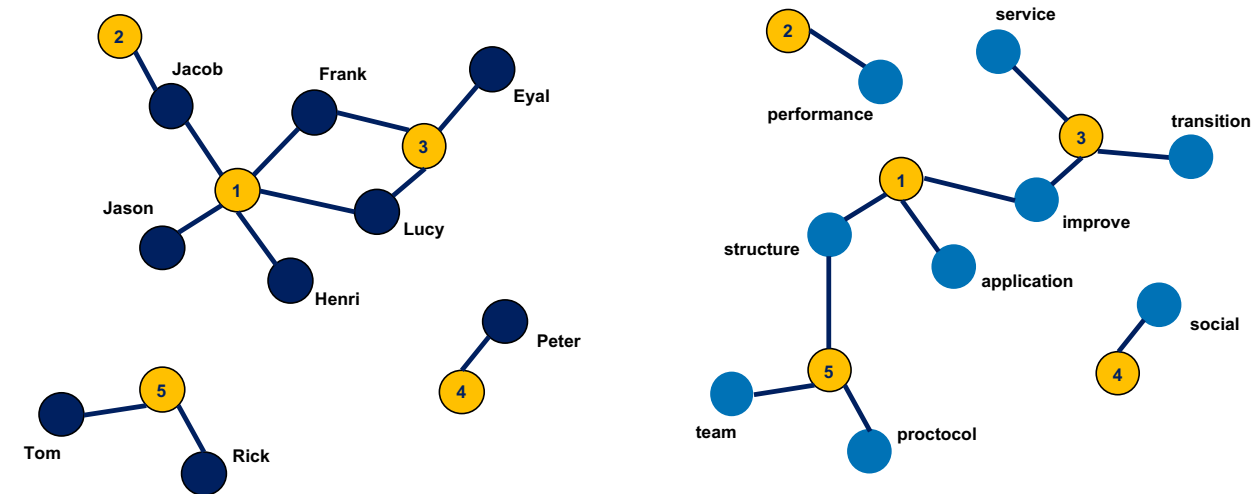
- Actors can be related to each other through the different symbolic forms they use, e.g. ideologies, as in a two-mode network of actors and symbols.



Ideas with impact:

How connectivity shapes idea diffusion

- A **positive interaction** between content and social network connectivity
- The highest diffusion success can be attributed to publications with **high content connectivity** and **high social connectivity**
- Ideas which **bridge** different knowledge domains in the content network will amass even **more citations** when they are developed by teams that are **highly connected** in the social network of co-authorship teams



The Structural Space

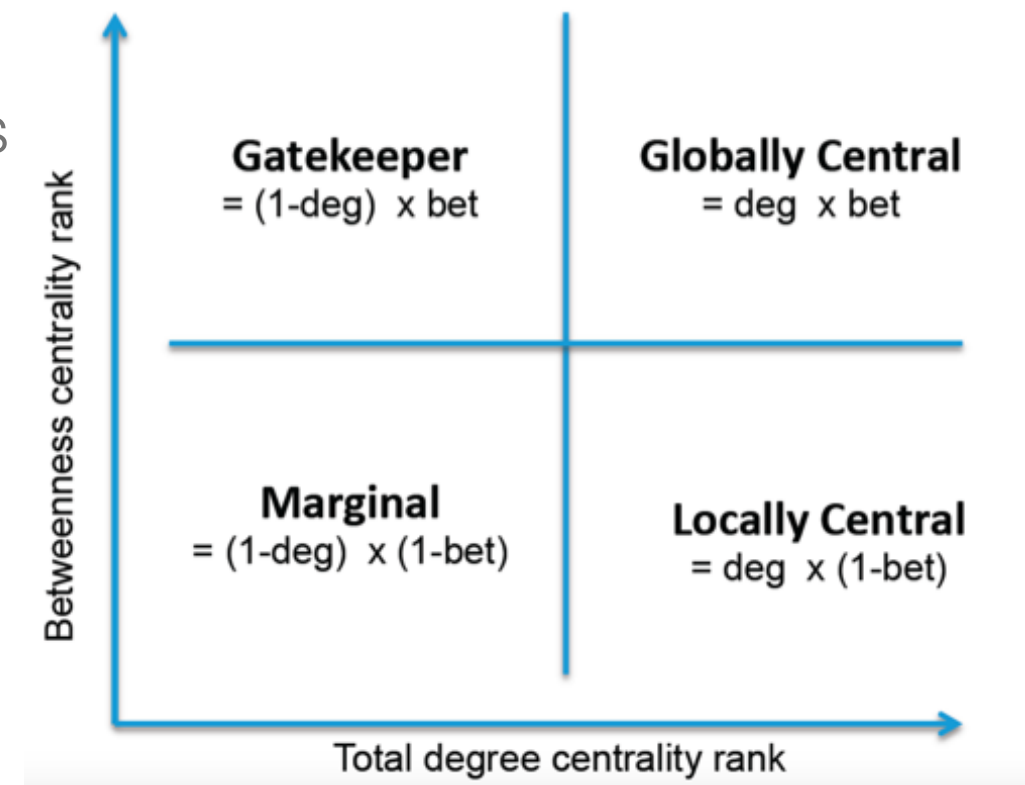
- The joint analysis of degree and betweenness centralities
- Useful in identifying those nodes that defy patterns often found in various network topologies

• Total Degree Centrality

- Local measure
- Popular concepts
- Important
- A hot topic's central key concept
- Able to activate many other key concepts

• Betweenness Centrality

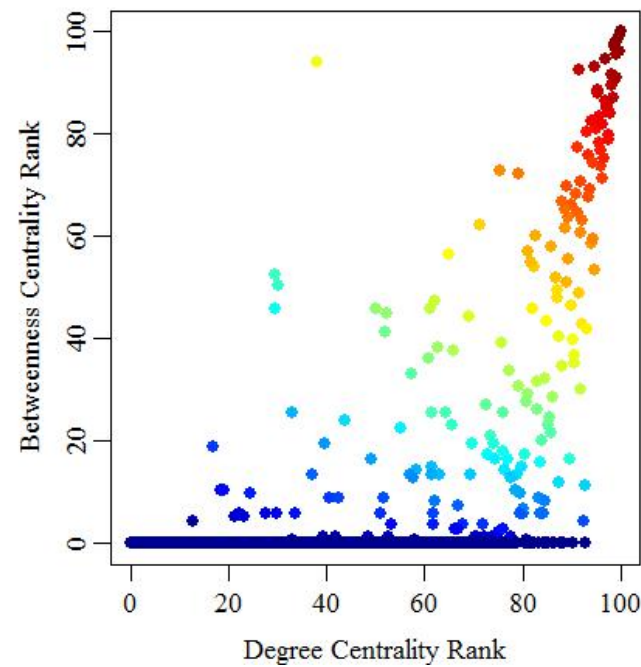
- Global measure
- Connective
- Influential
- Controls access to other key concepts
- Gatekeepers between different domains



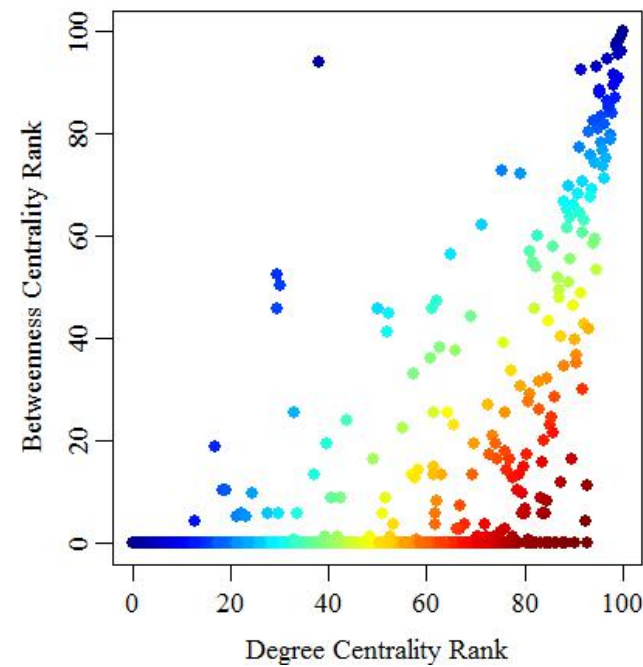
$$C_D^{\text{rank}}(i) = 100 \cdot \frac{j}{n} \mid (x_j^{\text{ordered}} = C_D(i))$$

4 Structural Roles

High popularity
+
High connectivity



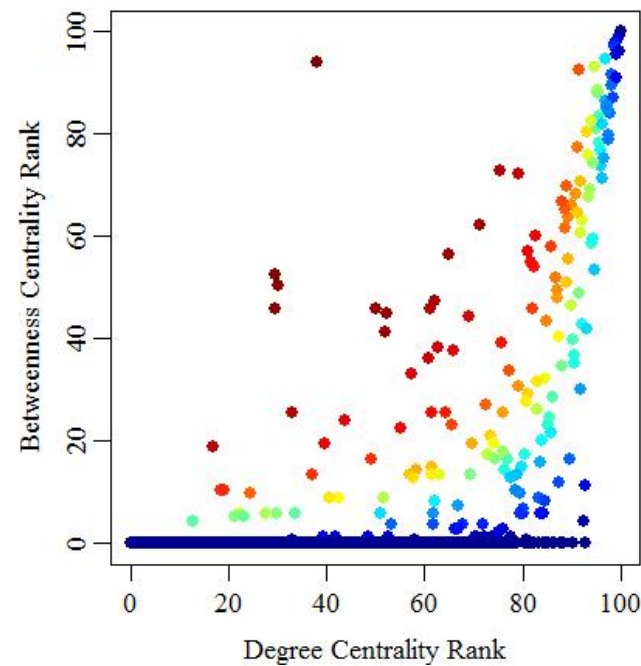
(a) Globally Central (GC)



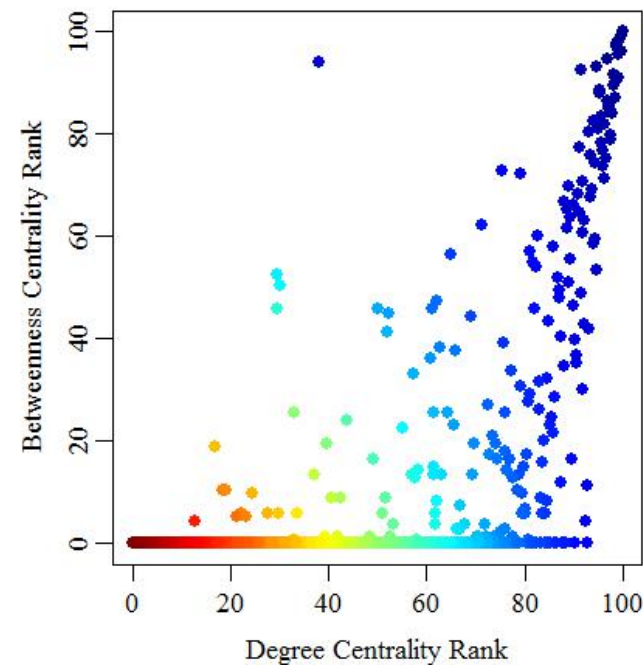
(b) Locally Central (LC)

High popularity
(local)

High connectivity



(c) Gatekeepers (G)

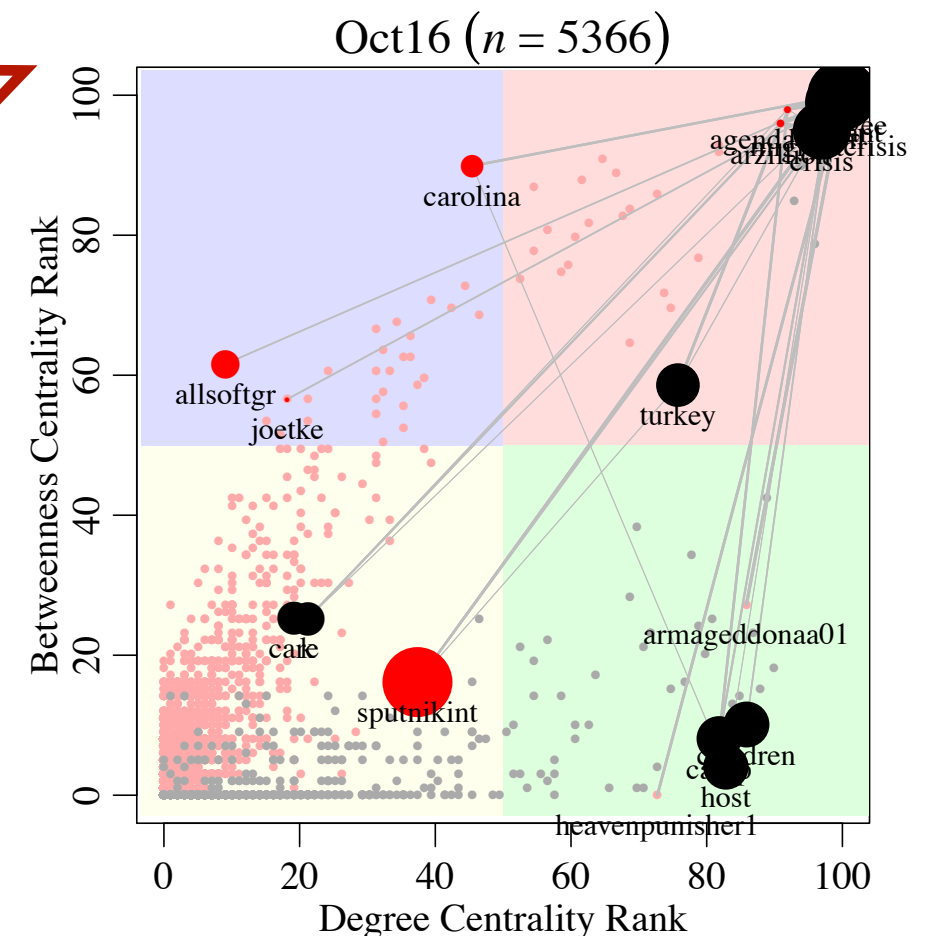
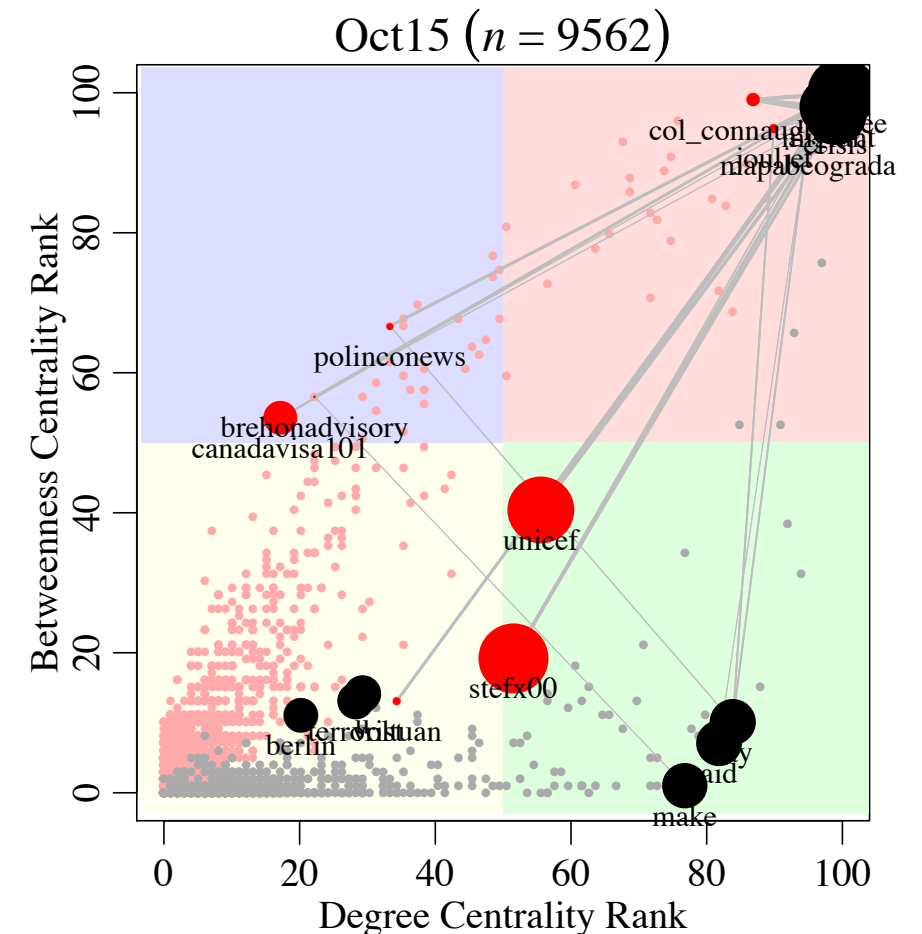
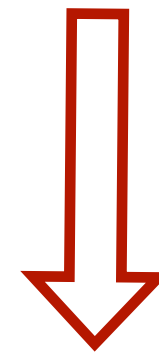


(d) Marginal (M)

Potential

Shifts in perspectives and positions on Twitter on the Refugee Crisis

- 'refugee', 'migrant', and 'crisis' at the center of the debate (GC) throughout
- Aid organizations enter the debate strategically positioned to span discussion communities
- By Oct16, charitable and issue-oriented organizations are replaced by 'self-proclaimed' activists
- Marginalized issues and individual enter the debate (refugee camps within Africa)



Thank you!

