

Outline of the talk

- 1 Basic concepts in network science
- 2 Applications to linguistic networks
- 3 Cognitive-linguistic applications: *one* example
- 4 Open questions

Basic concepts

What is 'network science'?

Definition

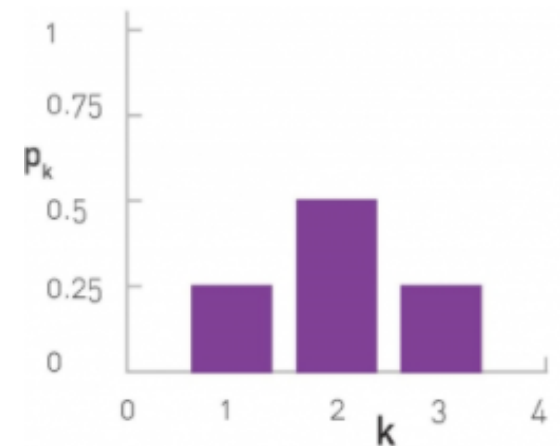
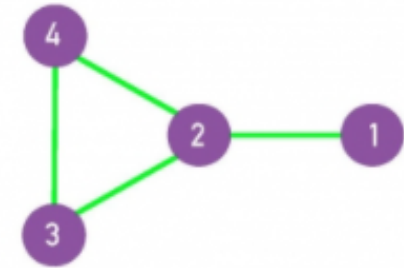
- The formal study of network structures across domains, using a common set of mathematical tools (based on Barabási 2016)
- Foundational papers: Watts & Strogatz (1998) on 'small world' networks; Barabási & Albert (1999) on 'scale-free' networks

Some major applications

- Sociology (e.g. networks of friendship or kinship ties)
- Technology (e.g. the internet)
- Economics (e.g. trade relations)
- Biology and health (e.g. genetic structures, virus spread)
- Neuroscience and artificial intelligence (neural networks)

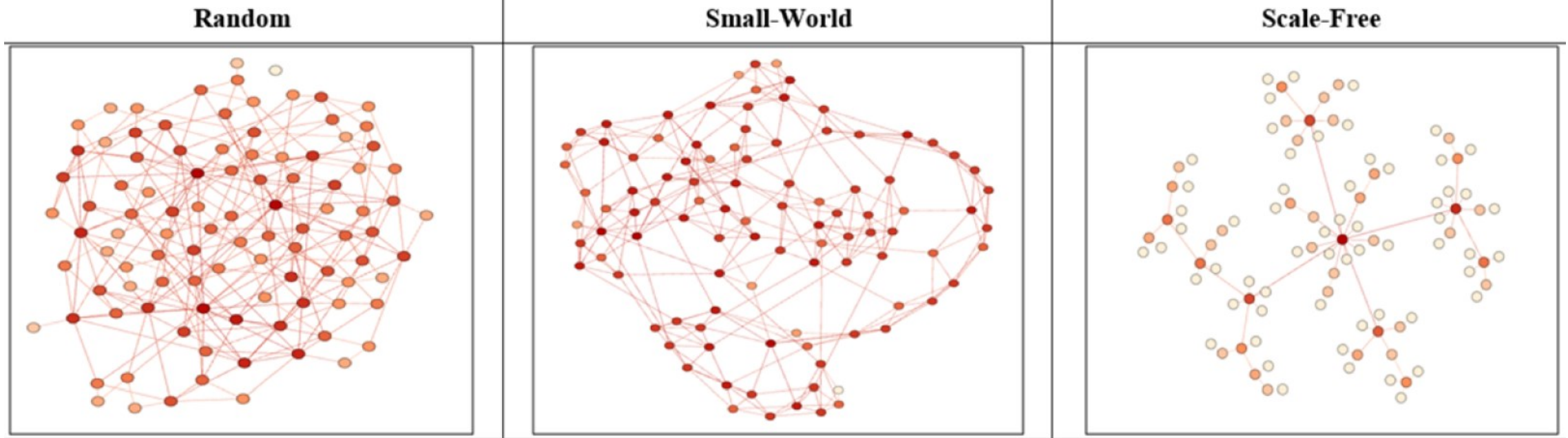
Basic concepts and network measures

- A network (= graph) consists of nodes (= vertices) and links (= edges)
- Types of networks: directed vs. undirected; weighted vs. unweighted; ...
- Network measures (local and global):
 - Degree k of a node (= how many links a node has)
 - Degree distribution $P(k)$ (= proportion of nodes with degree k)
 - Average shortest path length L (= how many steps from one node to another)
 - Clustering coefficient C (= probability that two neighbours of a node are linked to each other)
 - Modularity Q (= how many links are within network communities vs. between communities)



(Barabási 2016: Image 2.3)

Three topological structures



Short average path length

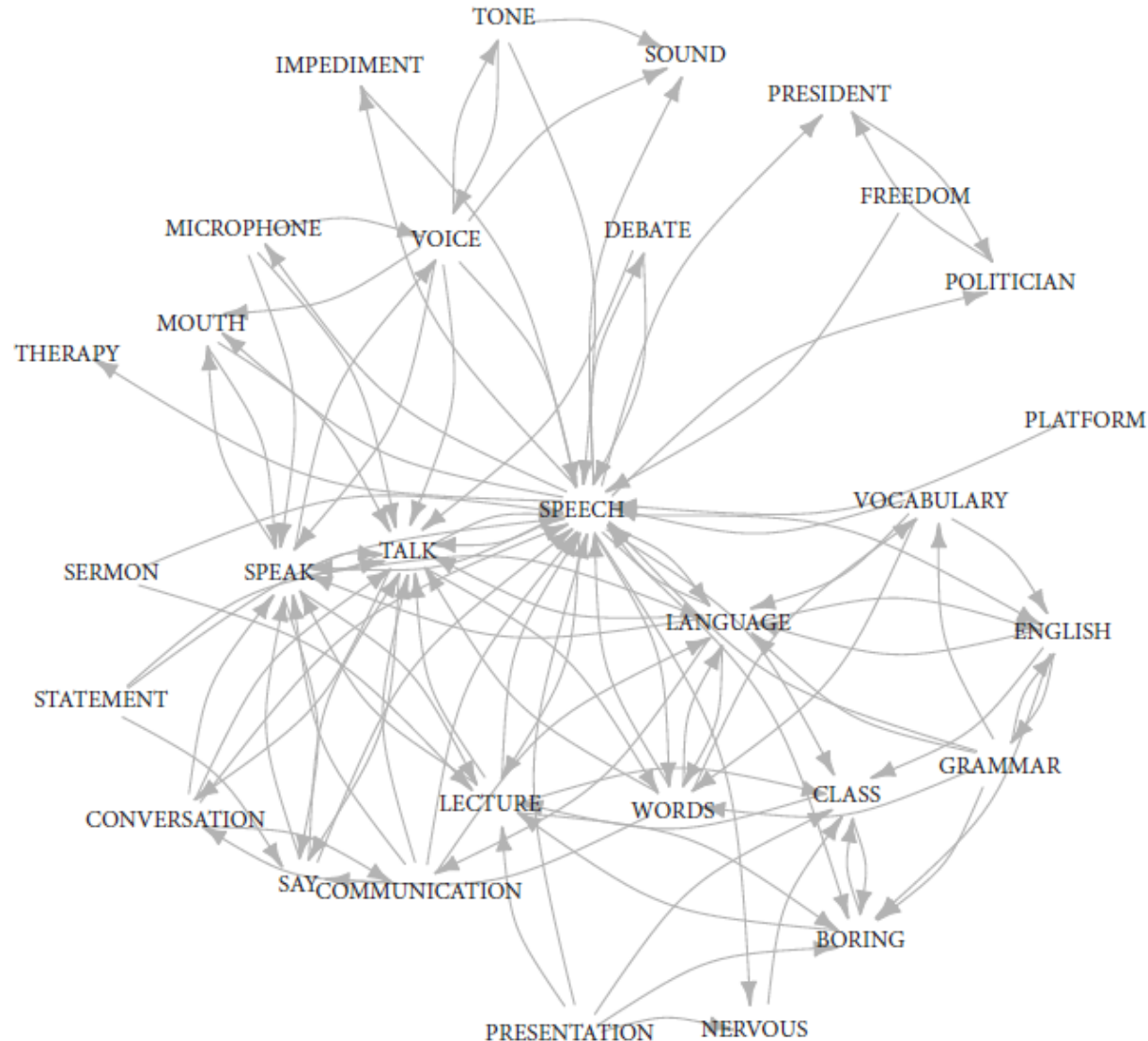
Short average path length,
large clustering coefficient

Some nodes with very high
degree ('hubs');
many nodes with small degree
(the degree distribution follows
a 'power law')

(Perera et al. 2017: Fig. 1)

Applications to linguistic networks

Lexical/semantic networks



Structure and growth

(Steyvers & Tenenbaum 2005)

- Semantic networks exhibit a small-world and scale-free structure
- They may grow by preferential attachment ('the rich get richer'): words with more links are acquired first

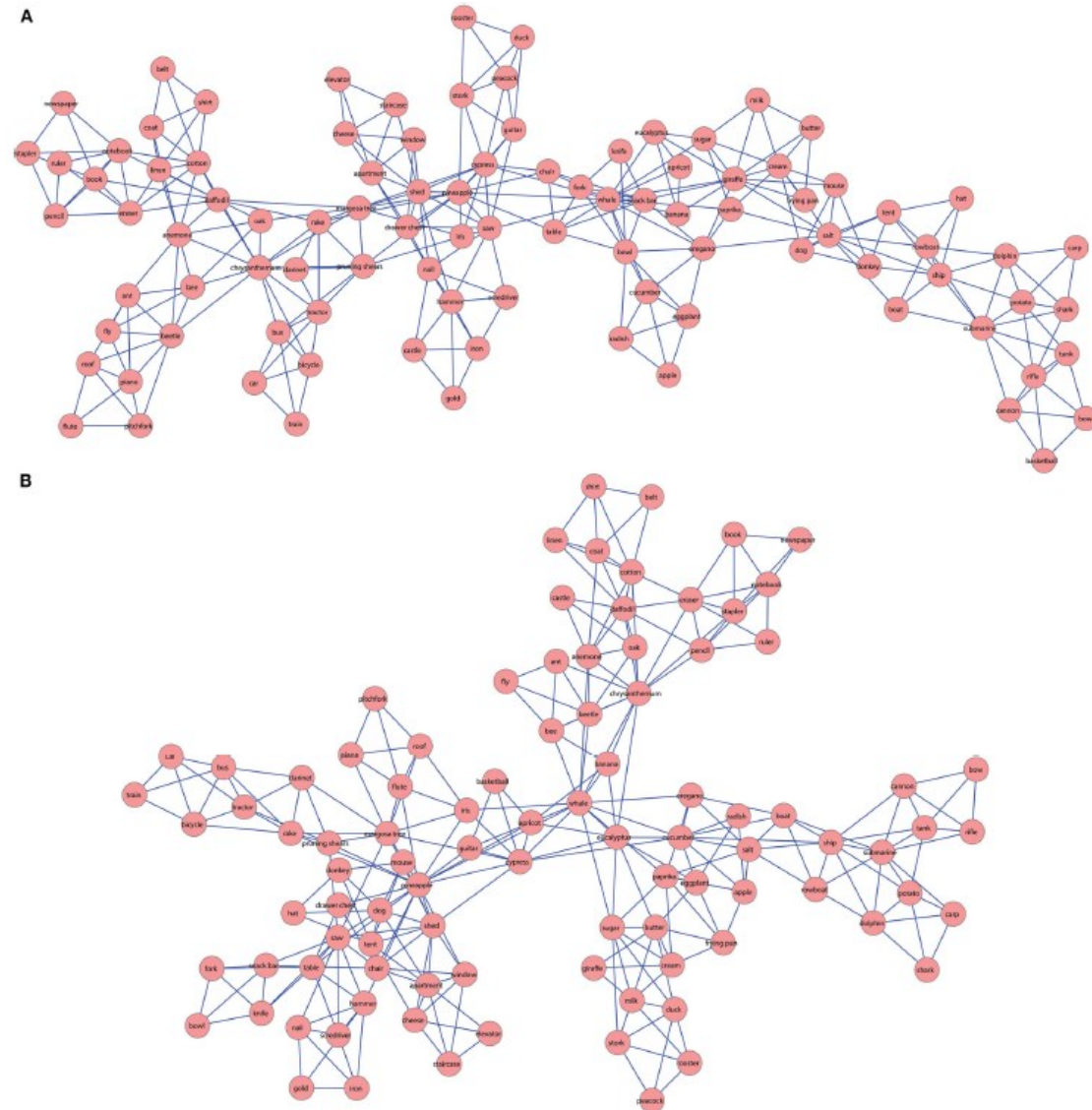
(Siew et al. 2019: Fig. 1)

Lexical/semantic networks

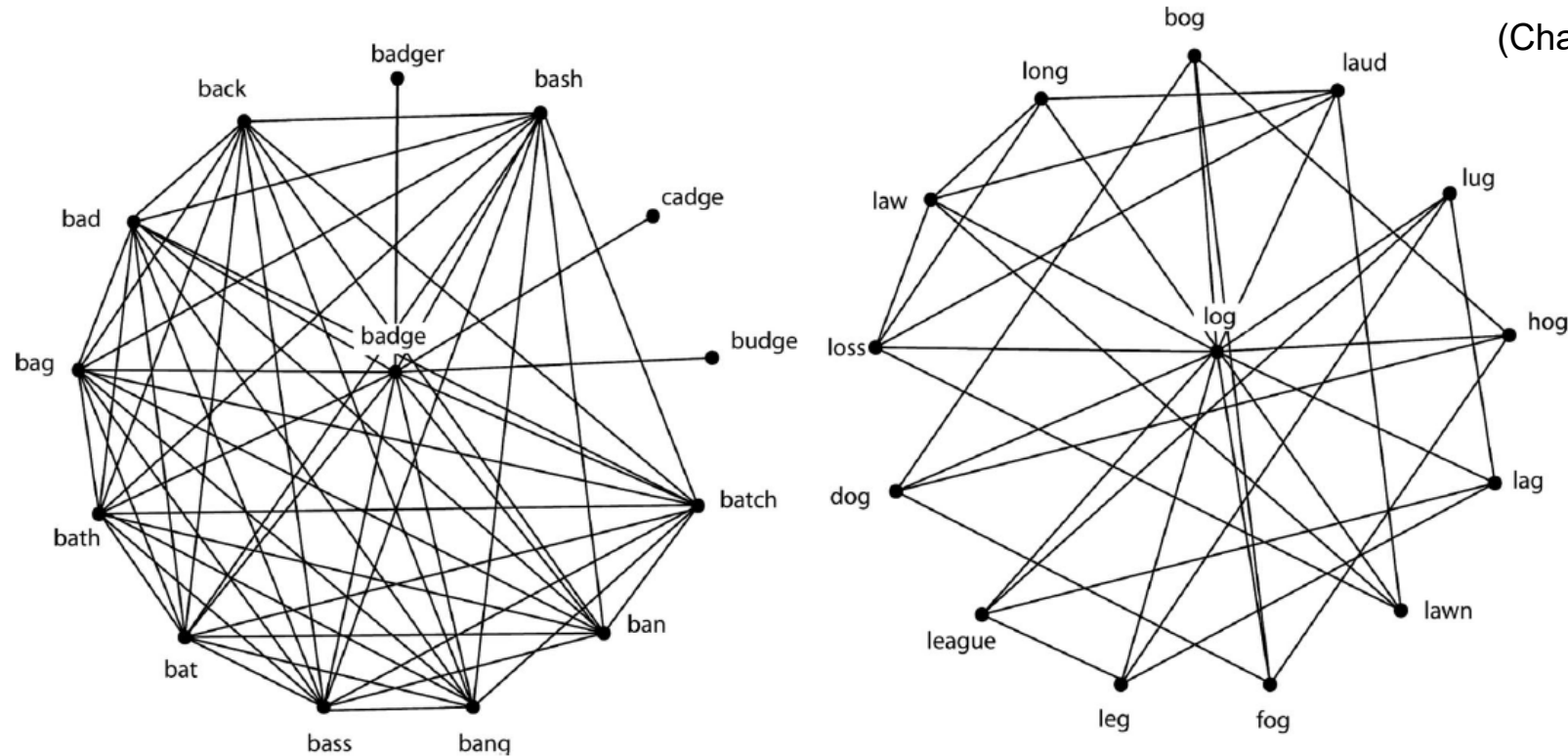
Creativity (Kenett et al. 2014)

- More creative persons have networks with higher clustering coefficient and smaller average path length (i.e. more 'small-worldness'), as well as lower modularity in community structure than less creative persons
- Creativity = flexible structures?

(Kenett et al. 2014: Fig. 1)



Phonological and orthographic networks



(Chan & Vitevich 2009: Fig. 2)

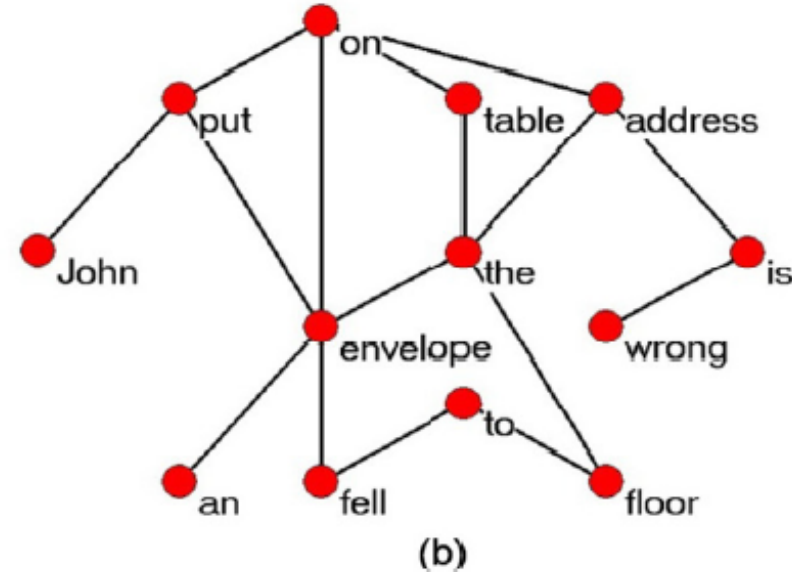
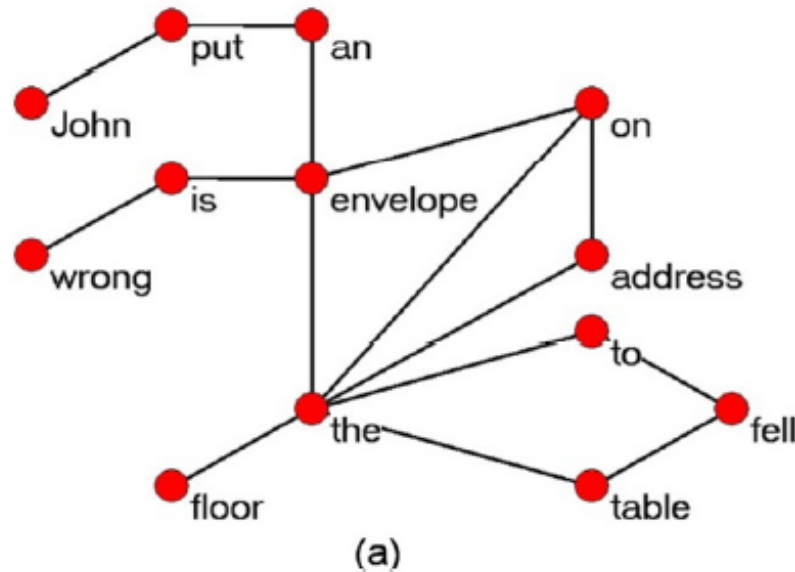
Predicting psycholinguistic effects (Chan & Vitevich 2009)

- Words with low clustering coefficient are recognised faster than words with high clustering coefficient
- The latter receive less activation because they have to 'share' with more neighbours?

Syntactic networks

Small-world and scale-free structure (Ferrer i Cancho & Solé 2001; Ferrer i Cancho et al. 2004)

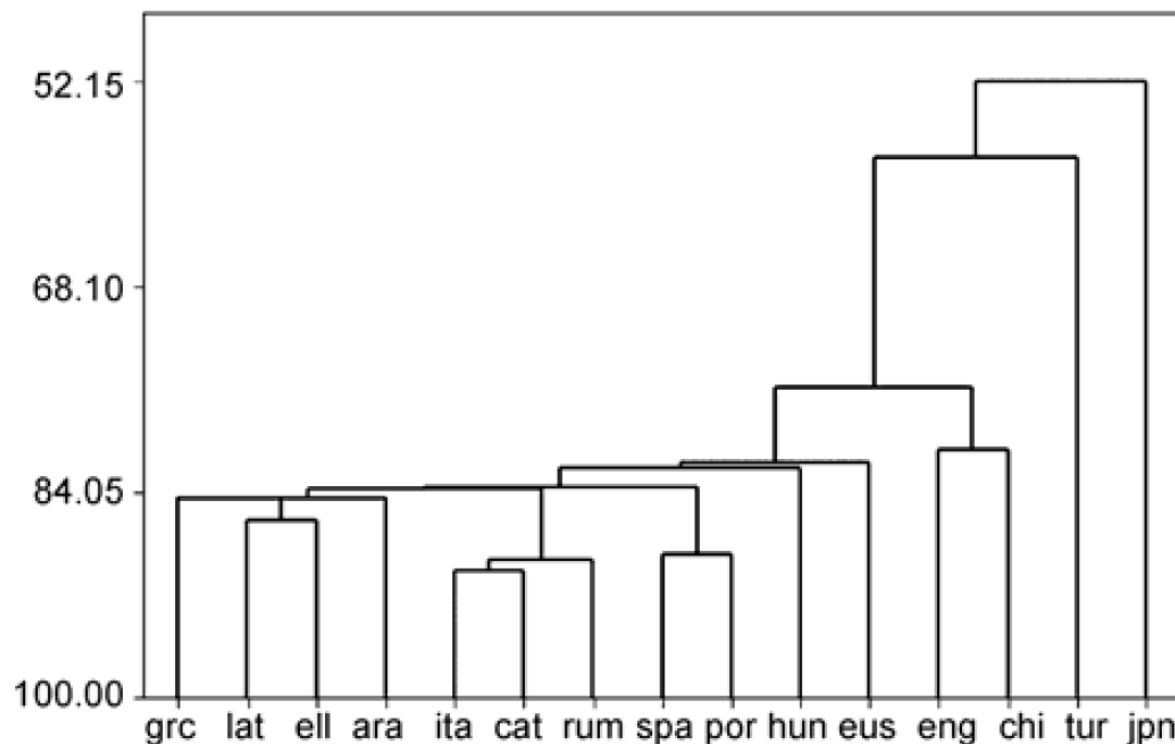
- ... are present in both word co-occurrence networks and syntactic dependency networks
- Network measures provide evidence of the hierarchical organisation inherent in dependency grammars



Syntactic networks

Language typology (Liu & Li 2010)

- Parameters of syntactic dependency networks (L , C , $\langle k \rangle$ etc.) can potentially predict typological similarity, even though the effect might be largely morphologically driven (Liu & Xu 2011)



(Liu & Li 2010: Fig. 6)

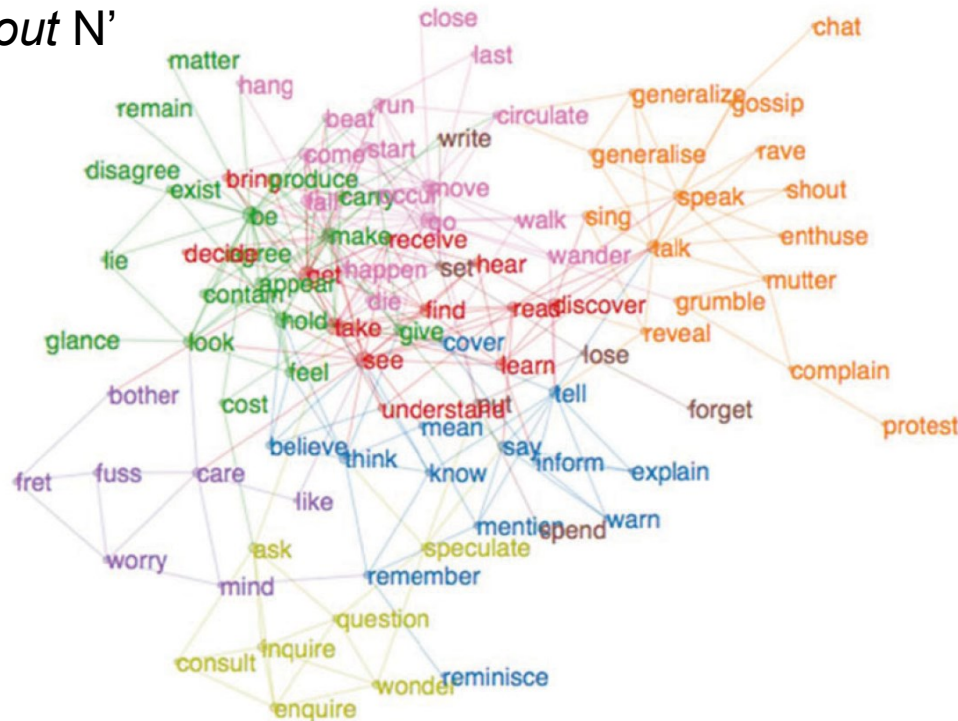
Cognitive-linguistic applications: *one* example

Ellis, Römer & O'Donnell (2016)

Semantic networks of verbs in verb-argument constructions (VACs)

- E.g. patterns like 'V *about* N', 'V *across* N', 'V *as* N'
- VACs are based on COBUILD patterns; similarity metric is based on WordNet database (Fellbaum 1998)
- Findings: semantic networks are well-connected (high C), have a few hubs (e.g. *say*, *see*, *go*) and form communities of related senses (e.g. communication expression, physical movement)

Ex.: 'V *about* N'



Psycholinguistic evidence

- The networks provide a measure of semantic prototypicality ('betweenness centrality'), which predicts how often L1 and L2 speakers generate the verbs in free association tasks
- Other factors are frequency and collocational attraction

(Ellis et al. 2016: Fig. 3.5)

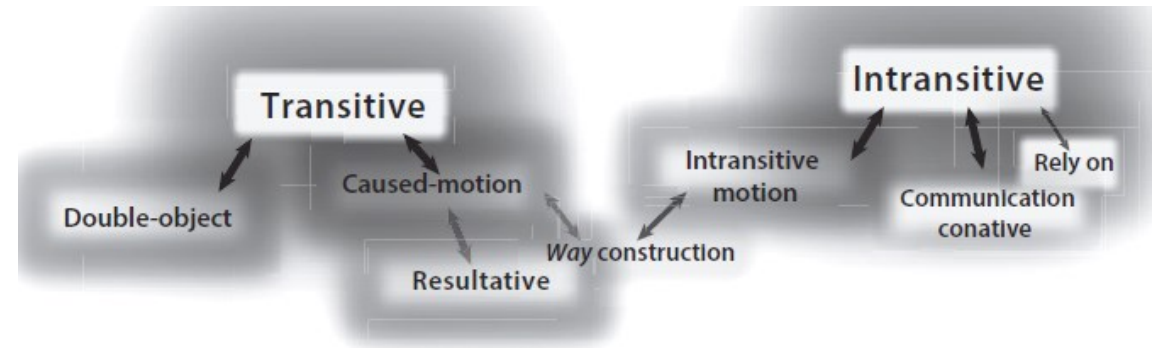
Open questions

Open questions

- 1) What other phenomena and questions in cognitive linguistics could be studied with the help of network science tools?
- 2) What kind of data would be required to construct the relevant networks?
- 3) To what extent can these methods be applied to Construction Grammar and Word Grammar networks?

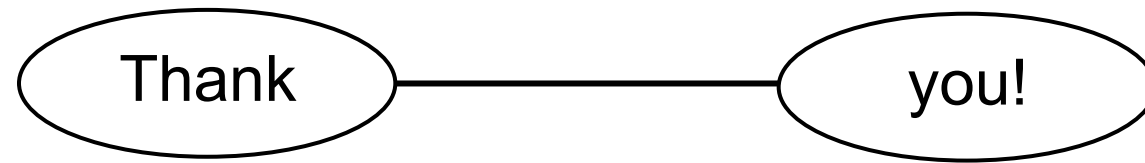
E.g. a question for CxG: what are the nodes and how many are there (words, low-level schemas, abstract phrase/clause-level cxns)?

- 4) Can Ellis et al.'s work be extended to compare verb usage across *multiple* constructions?



- 5) How much can network science measures tell us about psychological processes and representation? How cognitively plausible are the different types of networks (phonological, semantic, syntactic dependency networks)?

(Goldberg 2019: 37)



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