



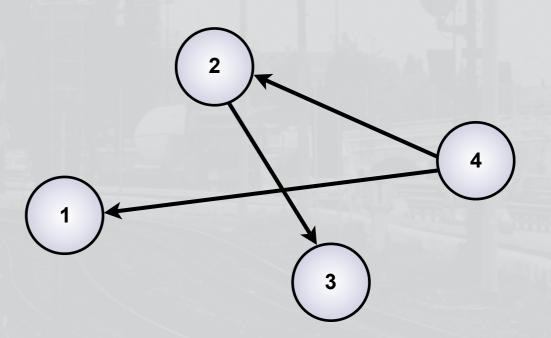
Overview

- What is a graph?
- Real problems as graph problems
- Graphs in neuroscience and neuroimaging
- Representing brain connectivity using graphs
- Advantages and limitations of this approach
- Robustness of in vivo connectomes
- The importance of subnetworks
- Combining information from multiple modalities
- Future work



Graphs

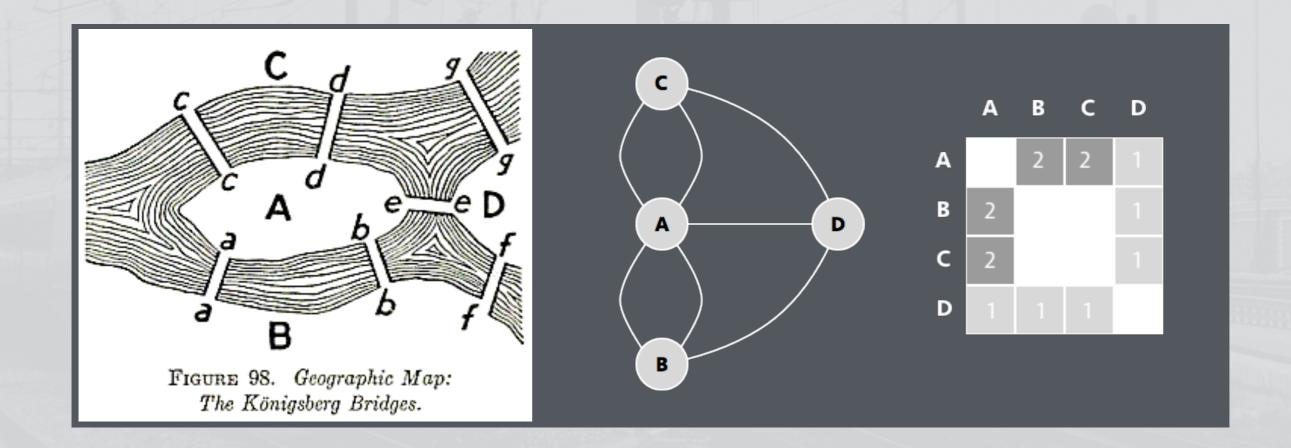
- A highly abstract representation of a set of vertices connected by edges
- Edges may be directed or undirected, and may have associated weights or costs
- A natural representation of connected systems
- Theoretically very well characterised
- Broad range of applications



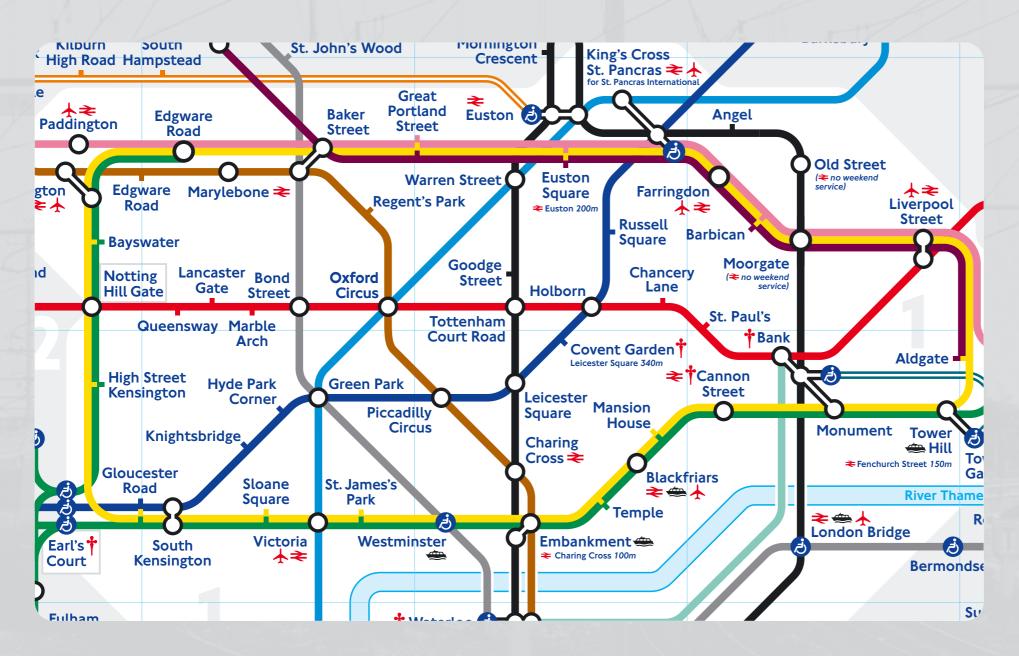


The bridges of Königsberg

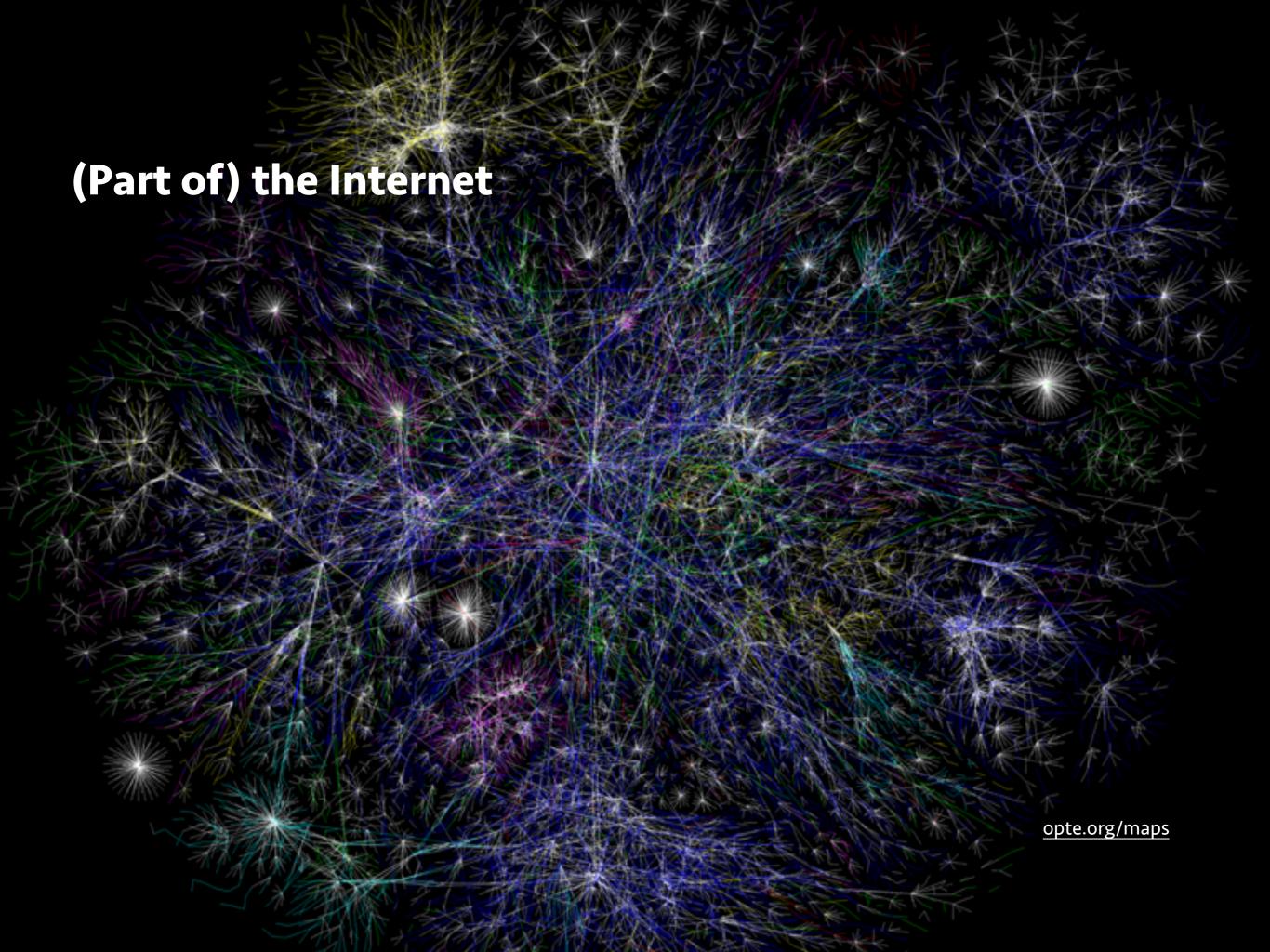
- Vertices are pieces of land; edges are bridges
- Can you walk around the town crossing every bridge once?
- Note: multiple links between two vertices make this a "multigraph"



A more modern graph problem



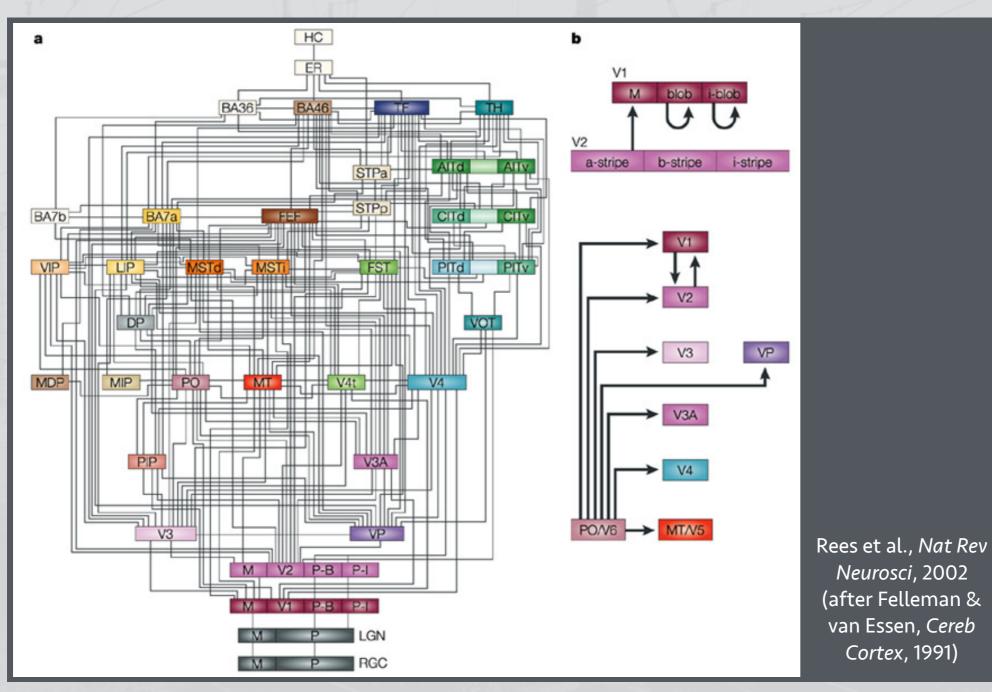
Travel times as edge costs





Connectivity and the brain

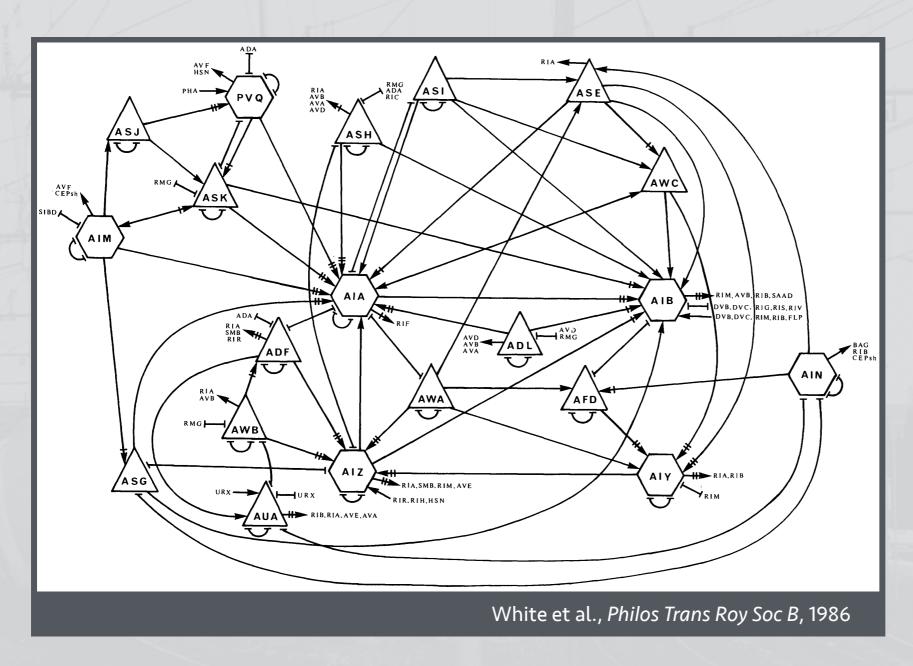
A/ I// WATER



Neurosci, 2002 (after Felleman & van Essen, Cereb Cortex, 1991)



The "connectome"



Circuitry of the nematode nervous system is fully mapped out



Connection and disconnection

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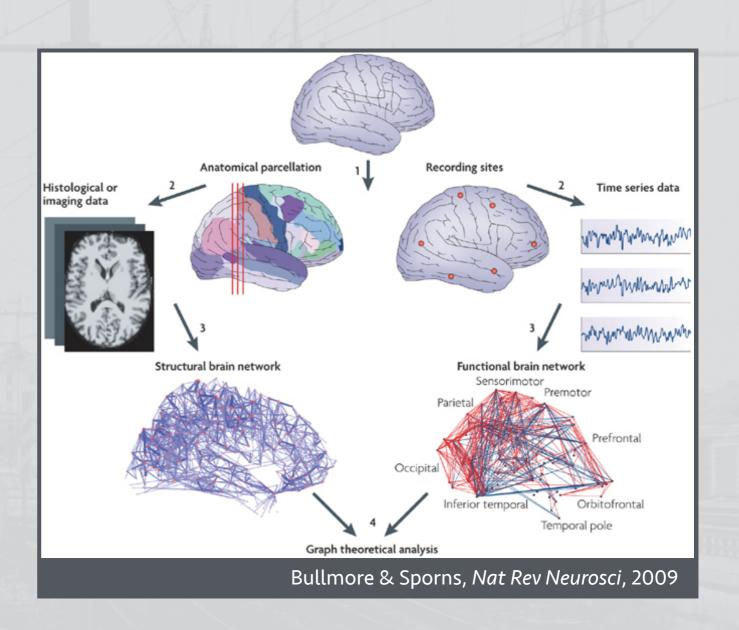
- Development of the brain's connectivity continues for years after birth
- Differences in connectivity patterns may underlie some of the variability in intelligence and cognitive information processing
- Disconnection between brain regions thought to be a key factor in agerelated cognitive decline
- Many neurological diseases are also thought to be associated with loss of connectivity (disconnection syndromes)
- Preserving connectivity is extremely important to ensure an optimal outcome after brain surgery
- Neuroimaging offers the chance to study connectivity in vivo
- Information can come from structural or functional MRI, EEG, MEG, etc.



Definitions of connectivity

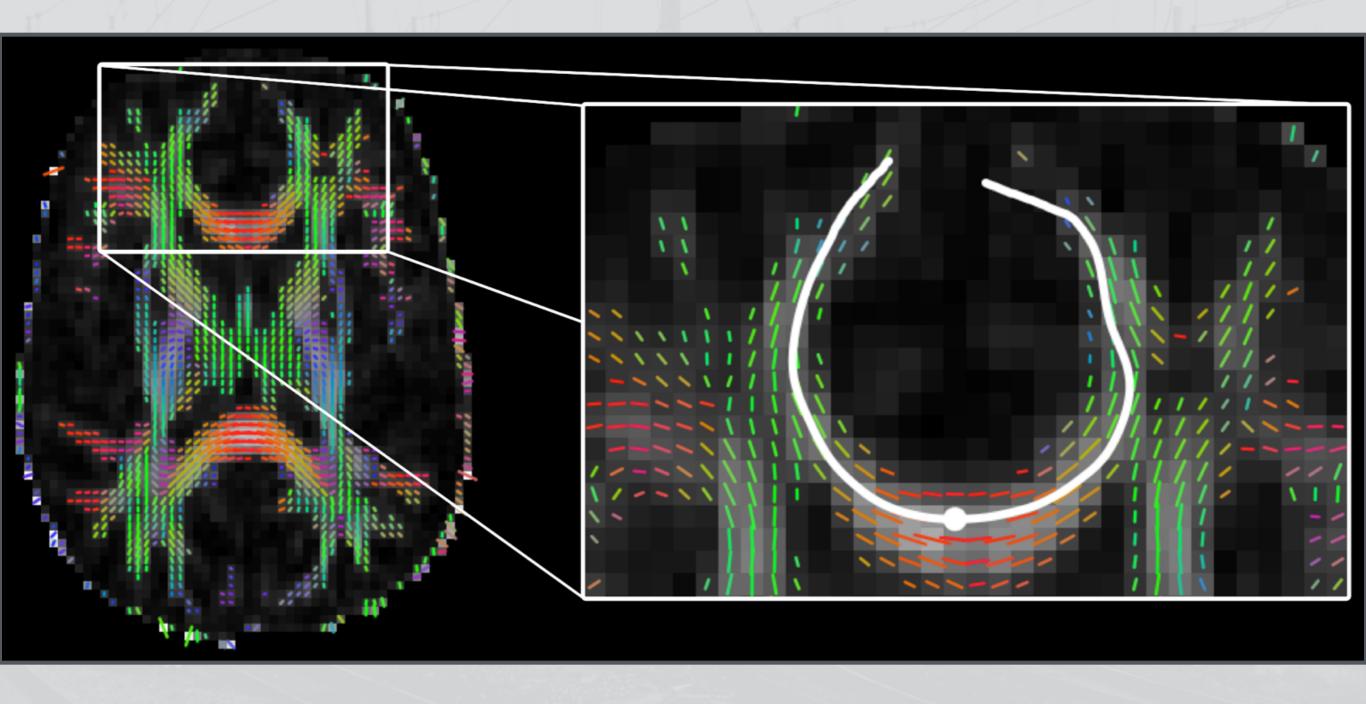
- Structural connectivity:

 the physical axon bundles
 connecting brain regions
 together
- Functional connectivity: associations between neural activity in spatially remote regions of grey matter
- Effective connectivity:
 patterns of influence by
 some neural systems over
 others



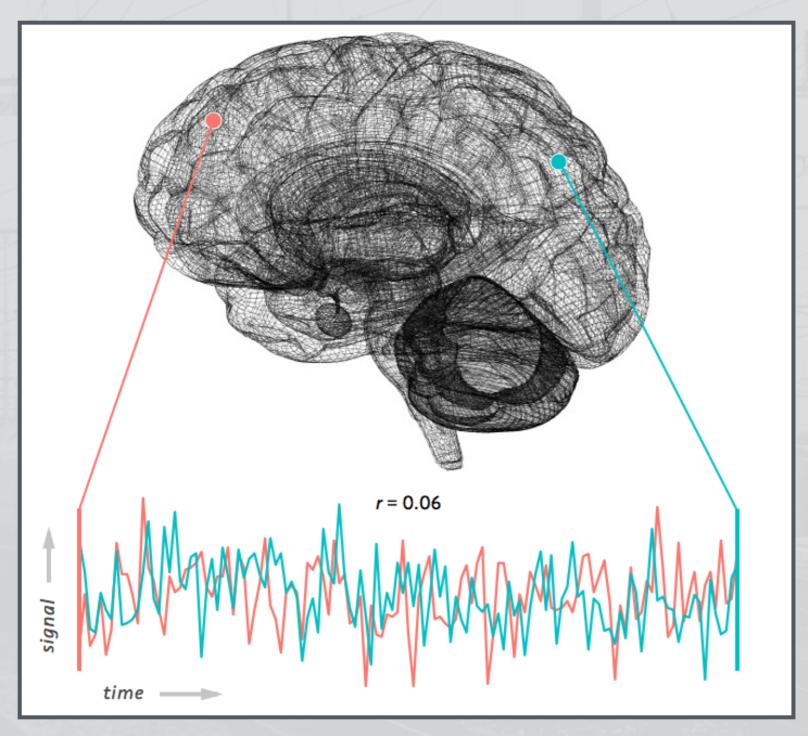


Structural connectivity: tractography



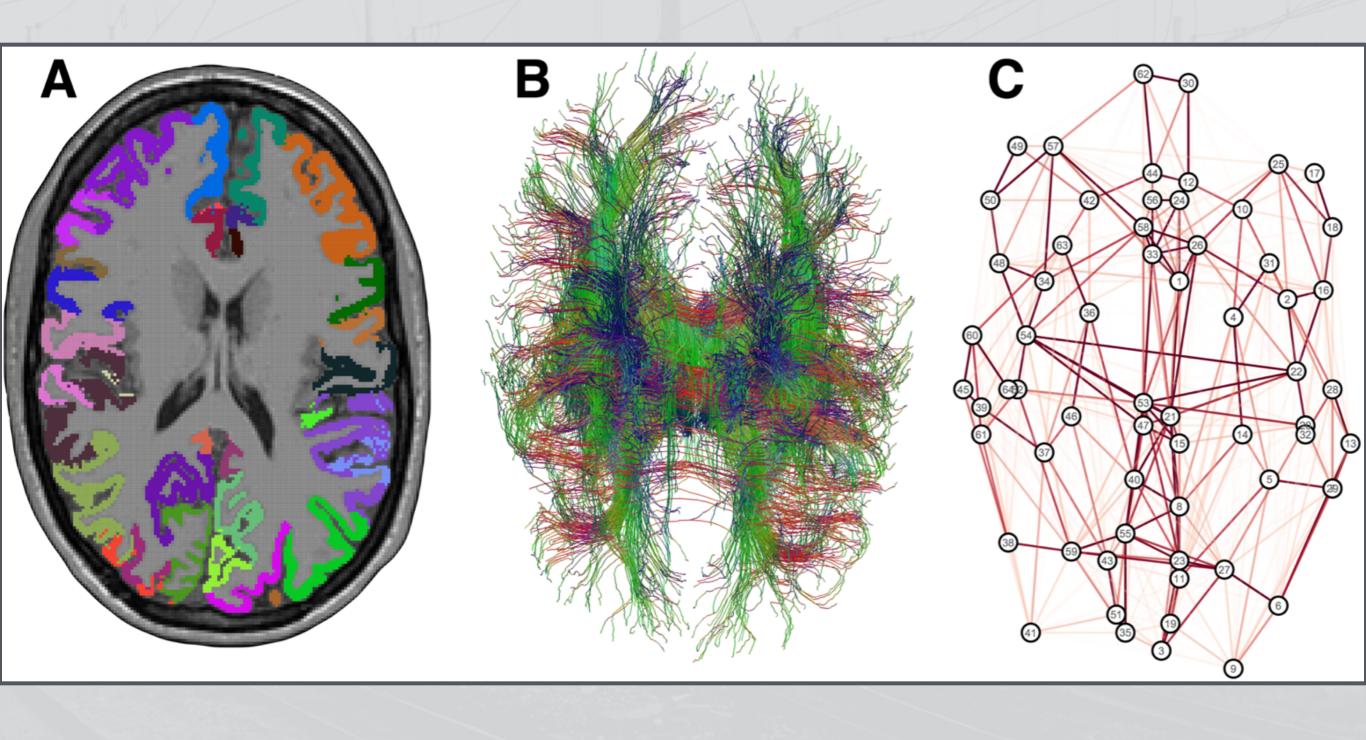


Functional connectivity: correlated time-courses





In vivo connectomics





Graph characteristics

- A range of measures have been developed in graph theory to describe characteristics of graphs and their vertices
- Connection density: the proportion of all possible edges which are present in the graph (cost)
- Average path length: the mean shortest path length between pairs of vertices (efficiency)
- Betweenness centrality: the number of shortest paths between other vertices which pass through a particular vertex (hubs)



Clinical and cognitive relationships

- Epilepsy patients show increased path lengths in cortical thickness networks (Bernhardt et al., Cereb Cortex, 2011)
- Changes to hubs and clustering properties of networks based on grey matter volume in patients with schizophrenia (Bassett et al., J Neurosci, 2008)
- Path length in functional networks related to intelligence (Langer et al., Hum Brain Mapp, 2011)
- Tractography-based structural network efficiency related to cognitive abilities in old age (Wen et al., *J Neurosci*, 2011)



Scope and limitations

- A graph can be created using any measure of association between brain regions of interest...
- ... but an association does not necessarily correspond to a direct connection
- Establishing direction of connections is challenging
- Abstract nature of graph makes systematic errors in underlying data invisible
- All the caveats of any preprocessing steps apply
- Choice of regions to use as vertices matters (cf. Zalesky et al., NeuroImage, 2010), but is usually arbitrary
- Substantial methodological variation in the literature



Some questions

- How robust are reconstructed connectomes?
- How can one identify important subnetworks without strong prior expectations?
- To what extent does structure predict function?
- How should one combine information from different modalities?

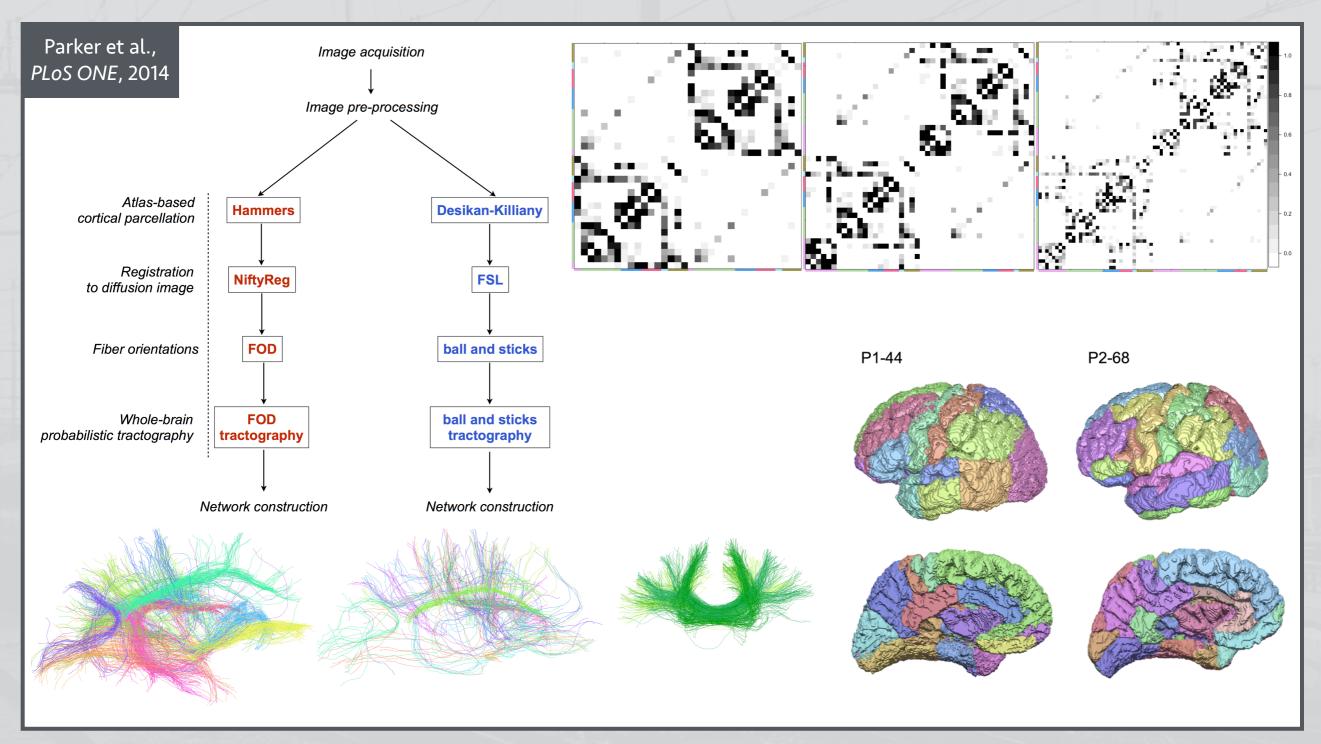


Robustness

- There is little consistency in the processes used to reconstruct connectomes
- Different pipelines may result in different results and therefore conclusions
- We need confidence in the robustness of the result if we want to make reliable scientific inferences



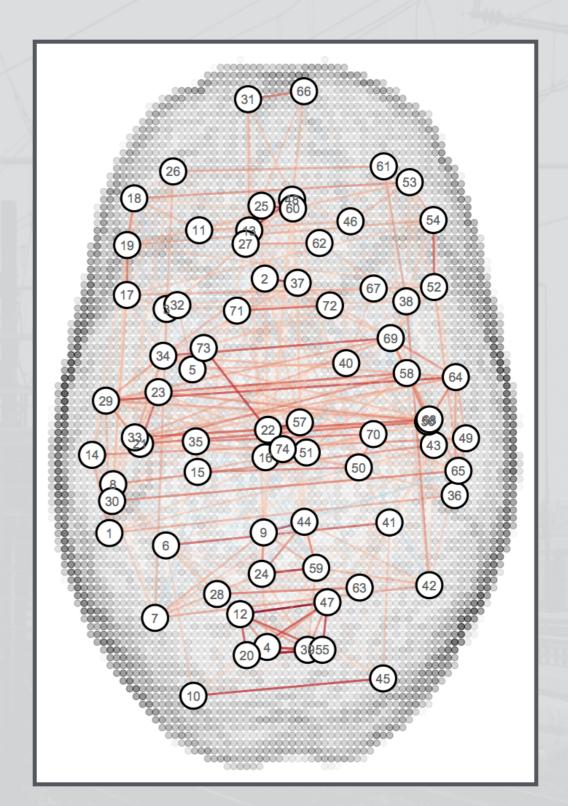
Robust connections across data processing pipelines





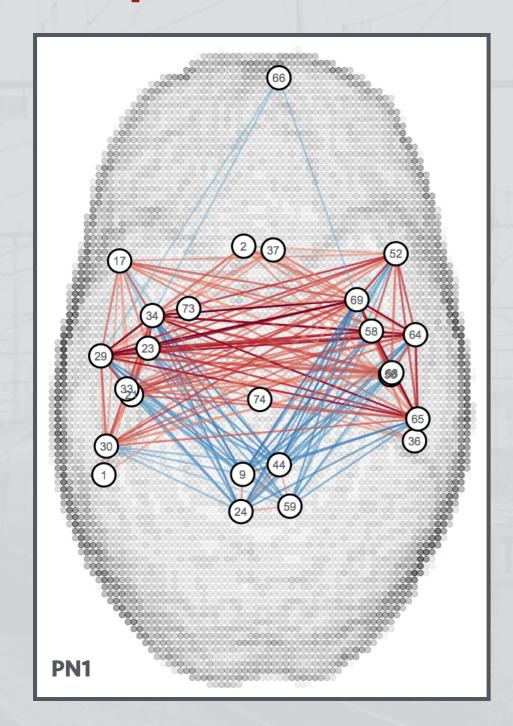
The whole connectome?

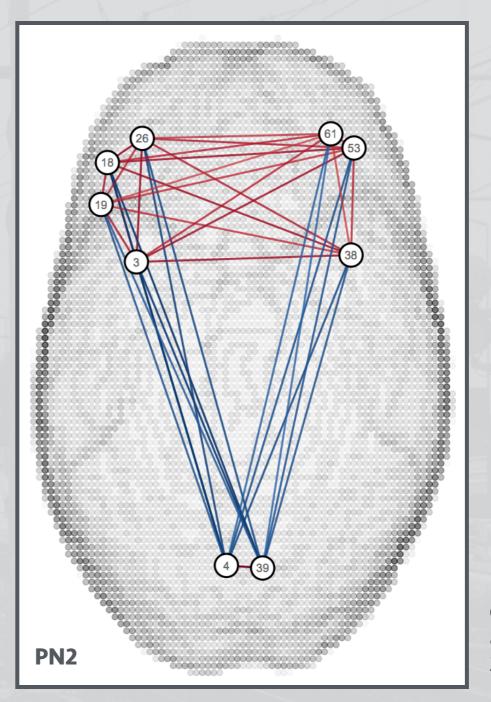
- Connectome-level analysis is now becoming common
- Graph theory can be used to describe network topology
- But it is reasonable to assume that the whole network is not involved in any given task
- Therefore development and disease processes may not show up as global topology changes
- Strategy: partition or decompose network into interesting subnetworks





Principal networks





cf. Clayden et al., *PLoS ONE*, 2013



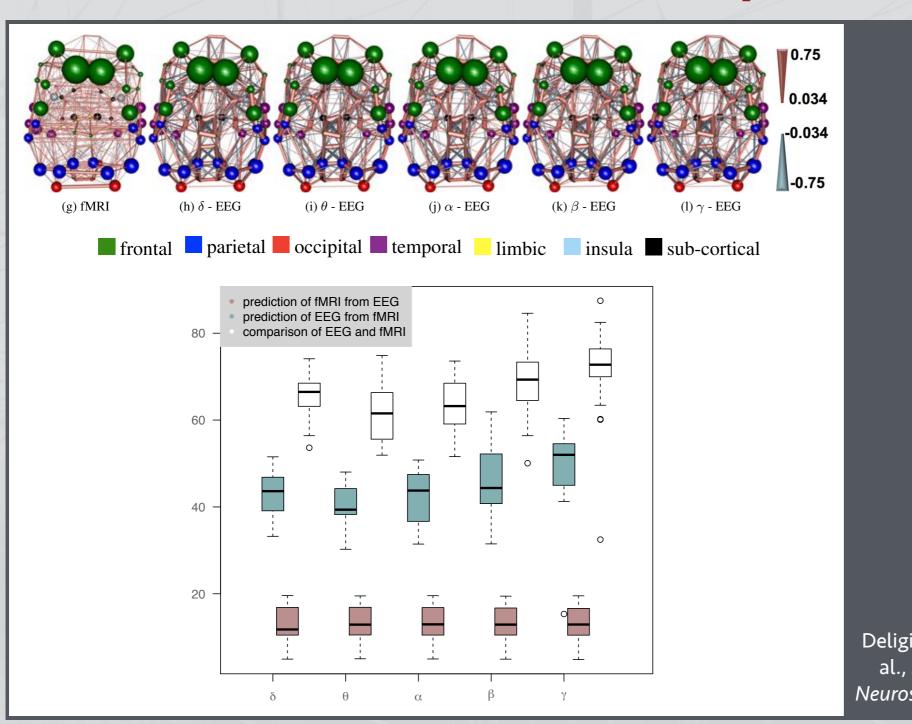
Between-modality prediction

- Given a connectome from one modality, can we predict another?
- Are some measures of connectivity better than others for this purpose?
- What information is most useful for the prediction?



Inter-modal functional connectivity

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Deligianni et al., Front Neurosci, 2014



Structural vs functional

- Structural and functional connectivity are conceptually different...
- ... but are underpinned by the same systems
- There should be substantial commonality
- Graph approaches allow the two to be represented similarly...
- ... but patterns of connectivity often differ in important ways
- Functional connectivity is often found to be more variable than structural connectivity
- Combining the two effectively remains an elusive goal for now



Areas of current and future interest

- Joint modelling of structural and functional connectivity
- Characterisation of population variability in connectivity patterns
- Integration of prior knowledge into connectome analysis
- Specialisation of image analysis approaches for sensitivity in particular diseases
- One-versus-many approaches for identifying abnormalities in individual patients



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