

#### A Tutorial in Connectome Analysis (I): Topological and Spatial Features of Brain Networks



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# Outline

- What are neural networks?
- Introduction to network analysis
- How can the fibre tract network structure be examined?
- Topological network organisation

#### What are neural networks?

#### Levels of connectivity







Axons between neurons

#### Links between cortical columns

Fibre tracts between brain areas

## Types of connectivity



- Structural / Anatomical (connection): two regions are connected by a fibre tract
- Functional (correlation): two regions are active at the same time
- Effective (causation): region A modulates activity in region B



#### **Cortical networks**



# Dorsal and ventral visual pathway





#### Introduction to network analysis

# **Network Science**

Rapidly expanding field: Watts & Strogatz, *Nature* (June 1998) Barabasi & Albert, *Science* (October 1999)

Modelling of SARS spreading over the airline network (Hufnagel, *PNAS*, 2004)

Identity and Search in Social Networks (Watts et al., *Science*, 2002)

The Large-Scale Organization of Metabolic Networks. (Jeong et al., *Nature*, 2000)

First textbook on brain connectivity (Sporns, 'Networks of the Brain', MIT Press, October 2010)

# Origin of graph theory: Leonhard Euler, 1736



Bridges over the river Pregel in Königsberg (now Kaliningrad) Euler tour: path that visits each edge and returns to the origin

# Nodes in graphs

- Isolated nodes
- Degree of a node
- Connected graph
- Average degree of a graph
- Edge density: probability that any two nodes are connected d= <u>E</u> (N\*N-1) /2)

- > Isolated node: v5
- Degree of a node:
   d(v1)=2, d(v4)=1
- Average degree of a graph: D = (2+2+2+1+0)/5 = 1.4



# Examples: edge density

	nodes	edges	density [%]
Autobahnen	1 168	2 486	0.18
Internet	6 524	29 629	0.0696
www	325 729	1 497 135	0.0014
Power Grid	4 677	12 500	0.0572
metabolic	422	1 972	1.3
C. Elegans	202	2 540	6.3
(partial netwo	rk)		
macaque	73	835	16

sparse network (density ~ 1%)

dense network
(density > 5%)

# How can the fibre tract network structure be examined?

# Tract tracing with dyes\*



PHA-L: Phaseolus vulgaris-leucoagglutinin

Anterograde:soRetrograde:so

soma  $\rightarrow$  synapse soma  $\leftarrow$  synapse

\* Horseradish peroxidase (HRP) method; fluorescent microspheres; Phaseolus vulgarisleucoagglutinin (PHA-L) method; Fluoro-Gold; Cholera B-toxin; Dil; tritiated amino acids

# Diffusion Tensor Imaging (DTI)





#### Topological network organisation

#### Archetypes of complex networks



Note: real complex networks show a combination of these types!

Kaiser (2011) Neuroimage

### It's a small world

**Nodes**: individuals

Links: social relationship





S. Milgram. Psychology Today (1967)



#### **Network properties**

<u>Clustering coefficient</u> Neighbours = nodes that are directly connected

local clustering coefficient  $C_{\text{local}}$ = average connectivity between neighbours

 $C_{\text{local}}$ =1 -> all neighbours are connected

C : global clustering coefficient (average over all nodes)

Characteristic path length

Shortest path between nodes *i* and *j*:  $L_{ij}$  = minimum number of connections to cross to go from one node to the other node

Characteristic path length L = average of shortest path lengths for all pairs of nodes



 $C_A = 4/10 = 0.4$ 



#### Small-world networks

Clustering coefficient is higher than in random networks

(e.g. 40% compared to 15% for the macaque monkey)

*Characteristic Path Length* is comparable to random networks



Small-world



#### Modular small-world connectivity



Small-world Neighbours are well connected; short characteristic path length (~2)

#### Modular

Clusters: relatively more connections within the cluster than between clusters

Hilgetag & Kaiser (2004) Neuroinformatics 2: 353

#### Hierarchy

#### Sequential

#### Topological





Spatial



Kaiser et al. (2010) Frontiers in Neuroinformatics Hilgetag & Kaiser PLoS Comput. Biol. (in preparation)

# Summary

# 2. Finding structural fibre tract connectivity:

- Diffusion tensor imaging
- Tract tracing

#### 3. Topological properties:

- multiple clusters/ modularity
- small-world: path lengths and local neighbourhood clustering

- **1. Types of connections:**
- Structural
- Functional
- Effective

#### **Further readings**



Jeff Hawkins with Sandra Blakeslee. On Intelligence. Henry Holt and Company, 2004



Olaf Sporns. Networks of the Brain. MIT Press, 2010



Duncan J. Watts. *Six Degrees: The Science of a Connected Age*. Norton & Company, 2004



Sporns, Chialvo, Kaiser, Hilgetag. Trends in Cognitive Sciences (September 2004) <u>www.dynamic-connectome.org</u>

## Practical

use Matlab or Octave

- Measures for brain connectivity structure and development (including data for the macaque and cat): <u>http://www.dynamic-connectome.org</u> <u>http://www.dynamic-connectome.org/t/tutorial/honey.mat</u>
- Brain Connectivity Toolbox: <a href="http://www.brain-connectivity-toolbox.net/">http://www.brain-connectivity-toolbox.net/</a>
- Connectome Viewer: <a href="http://www.connectomeviewer.org/">http://www.connectomeviewer.org/</a>

## Matlab analysis - topology

#### %% Network features using adjacency matrix *matrix*

%% see networks under the resources link at %% <u>http://www.dynamic-connectome.org/</u> %% for example, cat55.mat or mac95.mat

#### % how many nodes are there? N = length(matrix)

#### % how many edges are there (i.e. non-zero matrix elements)? E = nnz(matrix)

% what is the edge density (likelihood that any two nodes are connected? d = E / (N \* (N-1))

# % are there any loops (connections from node to itself)? min(min(matrix)) % any negative value out there? trace(matrix) % any non-zero diagonal elements (aka self-loops)

### Matlab analysis – spatial organisation

% network with 3D coordinates in variable *pos* e.g. using % <u>http://www.dynamic-connectome.org/t/tutorial/honey.mat</u>

#### %% visualize network

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spy(matrix) % binary view
pcolor(matrix) % view of values for weighted networks

hist(nonzeros(matrix))
unique(nonzeros(matrix))

#### %% Spatial Network visualisation

% view from top subplot(1,3,1); gplot(pos(:, [1,2])); axis equal

% view from side subplot(1,3,2); gplot(pos(:, [1,3])); axis equal

% view from back subplot(1,3,3); gplot(pos(:, [2,3])); axis equal