#### **Neuroinformatics**

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Week 6: Brain Connectivity Analysis (I)

# Outline

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

#### 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) cited 4,000+ times Barabasi & Albert, *Science* (October 1999) cited 4,000+ times

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

# Graphs

- Graph: set of nodes and edges (non-directed)
  G = (V,E)
- Set of nodes: V (singular: vertex; plural: vertices)
- Set of edges:  $E \subseteq V \times V$
- E.g., V={v1,v2,v3,v4}, E={(v1,v2), (v1,v3), (v2,v3), (v3,v4)}



# Directed graphs (Digraphs)

- Graph: set of nodes and arcs (directed)
- Set of nodes (vertices): V
- Set of edges:  $E \subseteq V \times V$ , the order matters
- E.g., V={v1,v2,v3,v4}, E={(v1,v2), (v1,v3), (v2,v3), (v3,v4), (v4,v1)}



## **Graphs and Networks**

In theory (mathematics) Graph: G=(V,E)

*Network*: N=(G, s, t, c)

defined by graph G with source s, sink t, and edge capacity c

(examples: electricity/power grid, water flow, metabolic flux)

In reality (CS, engineering, economics, life and social sciences): term network used throughout (as in this course)



# 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+3+1+0)/5 = 1.6



## 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: Retrograde: 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

### 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





Shortest path lengths: **A -> C** : 2 **A -> E** : 1

#### 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

Watts & Strogatz, Nature, 1998



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Small-world

#### Modular small-world connectivity



<u>Small-world</u> 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

#### Spatial network arrangement

#### Macaque (rhesus monkey) cortical network





## C. elegans neural network

#### Global level (277 neurons)



#### Most projections connect adjacent neurons



Kaiser et al. (2009) Cerebral Cortex

## Reducing neural wiring costs

- Minimizing total wire length reduces metabolic costs for connection establishment and signal propagation
- Component Placement Optimization, CPO Every alternative arrangement of network nodes will lead to a higher total wiring length



• Not the case! Reductions by 30-50% possible

### More long-distance projections than expected



## When long-distance fibers are missing (minimal):



Long-distance connections are short-cuts that help to reduce the characteristic path length (or average shortets path, ASP): Less long-distance connections -> longer path lengths Low path lengths = few intermediate processing steps are beneficial

- Synchrony of near and distant regions

- Reduced transmission delays

- Less (cross-modal) interference

- Higher reliability of transmission









#### Linking structure and function

### **Small-world properties**



EEG synchronization Network (functional connectivity)

in Alzheimer's patients and control group

Stam et al. Cerebral Cortex (2007)

#### Alzheimer: Path length vs. task performance

Mini Mental State Examination (attention, memory, language)



Diamonds: Alzheimer patients

Empty squares: Control

## Cognition: Path length vs. IQ

Resting state fMRI in 19 subjects (functional connectivity based on coherence)



Correlations between local path length lambda (shortest path length from one node to any other node) and IQ (Intelligence Quotient) for medial prefrontal cortex, bilateral inferior parietal cortex and precuneus / posterior cingulate regions (p < 0.05, corrected for age).

Van den Heuvel (2009) J. Neurosci. 29: 7619

## 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
  - 4. Spatial properties:
  - Preference for connections to neighbours
  - Fast processing due to long-distance connections

#### 5. Linking structure and function:

- Alzheimer's disease: path lengths
- IQ: path lengths

- 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>

#### Kaiser, Neuroimage, 2011