An introduction to brain networks

Alex Fornito

Brain and Mental Health Lab Monash Institute of Cognitive and Clinical Neurosciences Monash University

e: alex.fornito@monash.edu





Monash Institute of Cognitive and Clinical Neurosciences

two fundamental principles of brain organization

segregation









nanoscale

nm



connectome:

a comprehensive structural description of the network of elements and connections forming the human brain.

Sporns, Tononi, Kötter, PLoS Comp Biol, 2005

nm to µm

microscale

meso μ m to mm



macroscale mm to cm

modelling brain networks as graphs



any network can be modelled as a graph of nodes connected by edges nodes represent fundamental processing units edges represent the interactions between nodes

a brief history of graph theory







Leonhard Euler (1707-1783)



Paul Erdős (1913-1996)





Alfréd Rényi (1921-1970)

a brief history of graph theory



Barabasi & Albert, Science, 1999; Barabasi & Oltvai, Nat Rev Genet, 2004



Watts and Strogatz, Nature, 1999

real networks are modular



Girvan & Newman, PNAS, 2004; Clauset, Moore & Newman, Nature, 2008

real networks are small-world

defining nodes



chemoarchitecture



Eickhoff et al. NeuroImage, 2007



Brodmann, 1909

anatomical



Tzuorio-Mazoyer, et al. NeuroImage, 2002 Desikan, et al. NeuroImage, 2006



Area 4a

probabilistic

Area 1

Area 2

Area OP

Area TE 12

Hagmann, et al. PLoS One, 2007 Zalesky, et al. NeuroImage, 2010 Fornito et al. Font Sys Neurosci, 2010



functional

Area 17 Area 18

Dosenbach, et al. Science, 2010 Fornito et al. PNAS, 2012

data-driven



Yeo et al. J Neurophysiol, 2011 Craddock et al. Hum Brain Mapp, 2012 Power et al. Neuron, 2011

voxel-based



van den Heuvel et al. NeuroImage, 2009 Hayasaka & Laurienti, NeuroImage, 2010

myeloarchitecture

Area 44

Area 45



Glasser & Van Essen J Neurosci 2011

multimodal



Glasser et al. Nature, 2016

defining edges

Structural connectivity

The physical (anatomical) connections between brain regions Intrinsically directed Directionality cannot be resolved with MRI

Functional connectivity

A statistical dependence between spatially distinct neurophysiological signals Estimated at level of measured signals Can be directed or undirected

Effective connectivity

The influence that one neuronal system exerts over another

Usually model-based

Describes causal interactions at the neuronal level

Friston et al, 1994, Hum Brain Mapping; Fornito et al, 2015, Nat Rev Neurosci



mapping structural connectivity in humans with DWI









nistic

probabilistic





Johansen-Berg & Rushworth, Ann Rev Neurosci, 2014; Mori & Zhang, Neuron, 2008; Tournier et al. Magn Res Med, 2011; Isenberg, 2011; Thomas, et al. PNAS, 2014

measuring connection weights with diffusion MRI

connection weights typically estimated as: number of connecting streamlines

average tract FA or other diffusivity index

but:

NeuroImage 73 (2013) 239-254



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NeuroImage



journal homepage: www.elsevier.com/locate/ynimg

Comments and Controversies

White matter integrity, fiber count, and other fallacies: The do's and don'ts of diffusion MRI

Derek K. Jones^{a,b,*}, Thomas R. Knösche^c, Robert Turner^c

* Cardiff University Brain Research Imaging Centre (CUBRIC), School of Psychology, Gardiff University, Park Place, Cardiff, OF10 3AT, UK

^b Neuroscience and Mensal Health Research Institute, Cardiff University, Cardiff, CF10 3AT, UK

^c Max Planck Institute for Human Cognitive and Brain Sciences, Stephanstrasse 1A, 04013 Leipzig, Germany

emerging approaches include:

axon diameter & density (Assaf et al. NeuroImage, 2013) track density imaging (Calamante, et al. NeuroImage, 2011) myelin content (Deoni et al. Magn Res Med, 2008)



mapping functional connectivity with fMRI



Hutchison, et al. NeuroImage, 2013

topological biases of common functional connectivity measures



Fornito, Zalesky & Bullmore, 2016, Fundamentals of Brain Network Analysis



Zalesky, Fornito, Bullmore, NeuroImage, 2012

from data to graph



matrices and graphs are formally equivalent



which graph best represents the brain?



which graph best represents the brain?



from data to graph



from data to graph





Graphs provide useful models of brain networks

A unified framework for representing multiscale organization

There are different methods for defining brain nodes and edges Each has pros and cons; constrains interpretation of findings

Brain networks can be directed/undirected and weighted/unweighted Node/edge heterogeneity and dynamics can also be accomodated

Brain networks can be analysed at the level of connectivity or topology Effects can be mapped at individual edges, nodes, or globally

graph theory/network science

Newman (2010) Networks: An introduction. Newman (2003) SIAM Rev. Albert & Barabasi (2002) Rev Modern Physics graph theory and the brain Sporns (2011) Networks of the brain. Sporns (2012) Discovering the human connectome. Bullmore & Sporns (2009) Nat Rev Neurosci. Bullmore & Bassett (2011) Annu Rev Clin Psychol. Fornito, Zalesky & Breakspear (2013) NeuroImage.

software

Brain connectivity toolbox: https://sites.google.com/site/bctnet/

Graph analysis toolbox: https://www.nitrc.org/projects/gat/

Network-based statistic: http://www.nitrc.org/projects/nbs/

Task-related functional connectivity (cPPI): http://www.nitrc.org/projects/cppi_toolbox/

Network visualization http://immersive.erc.monash.edu.au/neuromarvl/

resources

Fundamentals of Brain Network Analysis Alex Fornito, Andrew Zalesky, and Edward Bullmore



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Elsevier http://bit.ly/1Hp29OZ

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