What actually is "connectivity"?

Stephen J. Gotts Laboratory of Brain and Cognition NIMH/NIH Bethesda, MD

In the context of non-invasive neuroimaging, "connectivity" is not really anatomical connectivity:

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It's a label that stands in for a set of (pretty complicated) measures that index anatomical and physiological proxies for actual synaptic connections

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For physiological measures, "connectivity" is constrained by anatomical connections but is not a mirror image of them (due to polysynaptic and network-level interactions)

A and C will often appear "connected" also

Examples:

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Structural Connectivity - DTI

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Functional Connectivity

- correlation(/regression) using BOLD EPI
- either in resting-state or task-based studies

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Structural Connectivity - DTI

Functional Connectivity

- correlation(/regression) using BOLD EPI
- either in resting-state or task-based studies

Effective Connectivity

- weight parameters within a causal model

Basic role of connectivity in brain functioning

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Spontaneous versus stimulus/task-driven activity across multiple levels of observation

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Connectivity measured with fMRI

Basic role of connectivity in brain functioning

Spontaneous versus stimulus/task-driven activity across multiple levels of observation

Connectivity measured with fMRI

How do we know that we're measuring what we want to?

Basic point:

Function comes from neurons, neurons activate each other via synaptic connections

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Each cortical neuron has a small (<1 mV) impact on any other, which means that *they must work together*

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But: Debate for decades over discrete stages vs interactivity

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Function comes from neurons, neurons activate each other via synaptic connections

But:

Debate for decades over discrete stages vs interactivity

$$
\longrightarrow
$$
 $Stage 1$ $Stage 2$ $Stage 2$

Early Views from Psychology and Cognitive Science: Discrete Stages and Modules (Marr, Fodor, etc.)

Basic point:

Function comes from neurons, neurons activate each other via synaptic connections

But: Debate for decades over discrete stages vs interactivity

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Function comes from neurons, neurons activate each other via synaptic connections

But: Debate for decades over discrete stages vs interactivity

Throughout cortex, many synaptic interactions are effectively bi-directional (e.g. V1<->V2, with Thalamus, etc.)

TRENDS in Neurosciences

Allene et al. (2015). Trends in Neurosciences 38, 524-34.

Cartoon of Laminar Structure in the Early Visual System

ascending pathways

(feedback)

Cartoon of Laminar Structure in the Rat Somatosensory System (Barrel Cortex)

Transition in thinking within the domain of language:

Poeppel et al. (2012). J Neurosci 32, 14125-31.

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Poeppel et al. (2012). J Neurosci 32, 14125-31.

Lichtheim/Geschwind Model

Wernicke's Aphasia

Transition in thinking within the domain of language:

Poeppel et al. (2012). J Neurosci 32, 14125-31.

Activity fluctuations and co-fluctuations

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Obviously not just limited to monosynaptic relations

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Obviously not just limited to monosynaptic relations

Sources of activity fluctuations:

Spontaneous activity Endogenous (i.e. internal, voluntary) Exogenous (i.e. stimulus-driven)

Activity (Co-)Fluctuations At Multiple Spatial Scales
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What we know from methods outside of fMRI:

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What we know from methods outside of fMRI:

Brain activity (action and synaptic potentials) never stops entirely in the absence of a stimulus.

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Brain activity (action and synaptic potentials) never stops entirely in the absence of a stimulus.

Cells in cortex typically fire at a baseline rate of \sim 1-10 spikes/second (Hz)

Spike recordings in monkey **Extrastriate Visual Cortex (V4)**: (e.g. Tolias et al., 2001, Neuron)

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Spike recordings in **Inferior Temporal Neurons**: (e.g. Desimone et al., 1984, J Neurosci)

Spike recordings in **Lateral Prefrontal Neurons**: (e.g. Rainer & Miller, 2000, Neuron)

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- spikes appear to have somewhat random times
- multiple presentations of identical stimuli don't produce identical spiking responses

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Conclusion for a long time:

Times of spikes don't matter, only average firing rate

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Conclusion for a long time:

Times of spikes don't matter, only average firing rate

However, with advent of multi-neuron recordings:

Spontaneous spikes are coordinated over large populations of cells

Spike/LFP recordings in **Primary Visual Cortex** using large electrode grid: (e.g. Kelly et al., 2010, J Comp Neurosci)

> Spontaneous activity 120 100 80

 (a)

Cell number

Spike/LFP recordings in **Primary Visual Cortex** using large electrode grid: (e.g. Kelly et al., 2010, J Comp Neurosci)

> (a) Spontaneous activity 120 100 Cell number 80 60 40 20 allah Vhi 78.A. käldes mandari ù. Ω

Local Field Potential (LFP)

Spikes

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2005 - Paul III (1996)

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Leopold et al. (2003). Cereb Cortex

LFPs are Coherent at Slow Frequencies Over both Short and Long Distances (e.g. *Schölvinck et al., 2010, PNAS*)

These fluctuations are reflected as synaptic currents in the LFP inside the head and (probably) serve as basis of EEG/MEG outside

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Slow, spontaneous fluctuations in spikes/LFP/BOLD occur in: Rest and Task

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Rest and Task Under Anesthesia

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Slow, spontaneous fluctuations in spikes/LFP/BOLD occur in:

Rest and Task Under Anesthesia In cortical slices removed from the brain Recordings from Primary Visual Cortex, *in vivo* (cat) and slice *in vitro* (ferret): (Sanchez-Vives & McCormick, 2000, Nat Neurosci)

Recordings from Rat Somatosensory cortex, *in vivo* and *in vitro* (slice culture): (Gireesh & Plenz, 2008, PNAS)

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Slow, spontaneous fluctuations in spikes/LFP/BOLD occur in:

Rest and Task Under Anesthesia In cortical slices removed from the brain

These fluctuations aren't noise, but are generated internally by the brain itself

Gregoriou, Gotts, Zhou, & Desimone (2009). Science 324, 1207-10.

Gregoriou, Gotts, Zhou, & Desimone (2009). Science 324, 1207-10.

Gregoriou, Gotts, Zhou, & Desimone (2009). Science 324, 1207-10.

100

100

80

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Gregoriou, Gotts, Zhou, & Desimone (2009). Science 324, 1207-10.

Gregoriou, Gotts, Zhou, & Desimone (2009). Science 324, 1207-10.

"Functional" Connectivity

"Effective" Connectivity

"Effective" Connectivity

Advantages of this type of study:

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Because done with electrodes and measuring action potentials, we're reasonably confident that what we're measuring is co-fluctuations of neural activity (not necessarily true in fMRI or even solely LFP/EEG/MEG)
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Disadvantages:

Really no better then fMRI and other methods in terms of inferring direct (monosynaptic) anatomical interactions

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Really no better then fMRI and other methods in terms of inferring direct (monosynaptic) anatomical interactions

Can't look at more than a handful of locations at once

Functional Connectivity of Spontaneous Activity at Rest (i.e. "Resting State")

- very popular (easy and fast to administer)
- subjects passively view a fixation cross
- fluctuations in spontaneous activity (< .1 Hz) are correlated throughout the brain in a spatially restricted manner

Functional Connectivity fMRI in Basic Research

• Cognitive, Systems, and Developmental Neuroscience

Functional Connectivity fMRI in Basic Research

• Cognitive, Systems, and Developmental Neuroscience

Functional Connectivity fMRI in Clinical Science

- Studying psychiatric disorders such as Autism and Schizophrenia, and Mood/Affective Disorders
- Neurodegenerative Disorders (PLS), Stroke, Neurosurgery

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In all cases, we'd like to separate neurogenic and artifactual sources of variation (but no perfect way of doing it)

Agreement of connectivity, task effects, and behavior is our best approach currently

fMRI Connectivity in Basic Research

Examples:

Parcellating the systems/circuit-level structure of functional interactions (e.g. Buckner and Petersen/Schlaggar labs)

Studying development (e.g. Fair)

Evaluating trait-like variation in behavioral abilities across subjects (Face Processing, Functional Lateralization)

Subgraphs change hierarchically over thresholds Spheres: areal, main cohort Surfaces: modified voxelwise, replication cohort

Power et al. (2011). Neuron Yeo et al. (2011). J Neurophysiol

7 Network Cluster Solution

Discovery Sample (n = 500)

Replication Sample (n = 500)

Power et al. (2011). Neuron

Alternative Parcellation Approach: Using Local Changes in Seed-based Correlation Maps

Nelson et al. (2010). Neuron

Boundaries generated using Canny method

placed over LLPC

Boundary map inversion to find peaks

Alternative Parcellation Approach: Using Local Changes in Seed-based Correlation Maps

Nelson et al. (2010). Neuron

Development of Functional Brain Networks

Fair et al. (2009). PLoS Comp Biol

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Resting-state Correlations Among Face-Selective Regions Predict Face Processing Ability Behaviorally

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Zhu et al. (2011). J Neurosci

- \overline{B} Face inversion \overline{C}
	-

Upright

Inverted

Global motion F

Whole-part

Whole

Part

Global-local

Inconsistent

Resting-state Correlations Among Face-Selective Regions Predict Face Processing Ability Behaviorally

Zhu et al. (2011). J Neurosci

Example from Our Lab: Functional Lateralization of Verbal, Visuospatial, and Motor Abilities

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Two distinct forms of functional lateralization in the human brain

Stephen J. Gotts^{a,1}, Hang Joon Jo^{b,1,2}, Gregory L. Wallace^a, Ziad S. Saad^b, Robert W. Cox^b, and Alex Martin^a

NAS

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Edited by Geoffrey K. Aguirre, University of Pennsylvania, Philadelphia, PA, and accepted by the Editorial Board July 25, 2013 (received for review February 8, 2013)

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PNAS PLUS

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Do the hemispheres differ in their within- vs between-hemisphere interactions ?

Does lateralization magnitude predict goodness of function?

Finding corresponding points in the two hemispheres:

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See also Jo et al. (2012). PLoS ONE

Comparing within- vs. between-hemisphere correlation at corresponding points:

Qualitatively Different Forms of Lateralization on Left vs Right

Qualitatively Different Forms of Lateralization on Left vs Right

Left-lateralized Effects (P<.005):

LL+LR > RR+RL ("Integration") LL-LR > RR-RL ("Segregation") Both

Right-lateralized Effects (P<.005):

RR+RL > LL+LR ("Integration") RR-RL > LL-LR ("Segregation")

Lateralization Magnitude Predicts Cognitive Ability

Lateralization Magnitude Predicts Cognitive Ability

Examples:

Examples:

Psychiatric Disorders: Autism, Schizophrenia, Bipolar Disorder

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Psychiatric Disorders: Autism, Schizophrenia, Bipolar Disorder

Neurological Disorders: Primary Lateral Sclerosis (PLS), Stroke

Autism (ASD) vs. Typically Developing (TD)
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Fractionation of social brain circuits in autism spectrum disorders

Stephen J. Gotts,¹ W. Kyle Simmons,² Lydia A. Milbury,¹ Gregory L. Wallace,¹ Robert W. Cox³ and Alex Martin¹

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Are the largest differences in functional connectivity concentrated among regions of the "social brain" ?

The "Social Brain" (a la Brothers, 1990; Frith & Frith, 2007; Adolphs, 2009)

Empirical Determination of "Seeds":

 $Z = -11$

Whole-brain Differences in Functional Connectivity: TD > ASD

Whole-brain Differences in Functional Connectivity: TD > ASD

Agreement with Social Symptom Correlations (ASD only)

Applying the same method to Childhood Onset Schizophrenia (vs. Typ. Developing)

Collaboration (SG, AM) with: Becky Berman Harrison McAdams Nitin Gogtay Judy Rapoport

Berman et al. (2016). Brain 139, 276-91

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2 Clusters of Regions with reduced correlation in COS relative to TD:

Social/Cognitive vs. Somatosensory/Motor

 $TD - COS$ $(df = 43)$ p<.05 5 (corrected) $\overline{\mathbf{0}}$ -5 p<.05 (corrected)

t-val

Group Differences Are Associated with Positive/Negative Symptoms in COS

Correlation with Positive Symptoms (SAPS)

Correlation with Negative Symptoms (SANS)

Increased Resting Correlations in Primary Lateral Sclerosis (PLS)

Collaboration with Mary Kay Floeter (NINDS) and Avner Meoded (Johns Hopkins):

Increased Resting Correlations in Primary Lateral Sclerosis (PLS)

Collaboration with Mary Kay Floeter (NINDS) and Avner Meoded (Johns Hopkins):

Controls PLS PLS-Control

5 Patients with Left Hemisphere Lesions due to left CVA

Collaboration with Laurel Buxbaum and Christine Watson (Moss Rehab Research Institute):

5 Patients with Left Hemisphere Lesions due to left CVA

Collaboration with Laurel Buxbaum and Christine Watson (Moss Rehab Research Institute):

Example Patient

MPRAGE

Example Patient MPRAGE EPI

Example Patient MPRAGE

EPI

Example Patient

MPRAGE

Tissue Masks

Lesion Reconstruction

Example Patient (After Data Cleaning)

R Postcentral Gyrus Seed (r > +0.35):

6mm-Radius Sphere

X = +47

R Intraparietal Sulcus Seed (r > +0.35):

 $X = -44$ **Z** = $+32$

Example Patient (After Data Cleaning)

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How do we find unusual correlation levels over the entire brain? (more systematically)

1) Find the correlation of every voxel with the RH voxels

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2) then average OR threshold above a certain value (e.g. > 0.2 or 0.3; as in Buckner et al., 2009, J Neurosci),

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- 2) then average OR threshold above a certain value (e.g. > 0.2 or 0.3; as in Buckner et al., 2009, J Neurosci),
- 3) store the average OR voxel counts (> thresh) back in each voxel

Average r-value $(-.15 < r < .15)$

log(# RH voxels > threshold)

Average r-value $(-.15 < r < .15)$ threshold = .2

log(# RH voxels > threshold)

Average r-value $(-.15 < r < .15)$ threshold = .2

log(# RH voxels > threshold)

threshold $= .3$

Example Patient < Penn Controls (p<.05, corrected)

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Summary

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• "Connectivity" reflects ongoing activity fluctuations and cofluctuations

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- fMRI Connectivity can be used to map large scale brain organization
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- "Connectivity" reflects ongoing activity fluctuations and cofluctuations
- fMRI Connectivity can be used to map large scale brain organization
- Individual variation in cognitive abilities (e.g. verbal, visual) and in patient symptoms is also reflected in connectivity measures
- Each of these phenomena demonstrates not only *reliability* of resting-state correlations, but *validity -* and they are most likely based in real neural covariation