# What actually is "connectivity"?

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





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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

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*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

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

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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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#### Lichtheim/Geschwind Model



Wernicke's Aphasia

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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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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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Cells in cortex typically fire at a baseline rate of ~ 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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- 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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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)

**(a)** Spontaneous activity 120 100 Cell number 80 60 40 67 20 -พ.ศ. . เพื่อ at the second se 2.67 . Na kata 0



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 (A,b)Na py a anna in 0

Local Field Potential (LFP)

**Spikes** 



5

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44

17

17. The second

4

Gratings

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147 (2014) 1990 - Star

34.

6

Time (s)

2.17

1 C. Martin

4.00.02

8



#### Natural movie



2

att

2

#### 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)



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 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.

### "Functional" Connectivity



### "Effective" Connectivity

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### Advantages of this type of study:

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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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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

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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)

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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)

#### Power et al. (2011). Neuron



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



#### 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



Square patch of 729 foci (27x27 grid) placed over LLPC



Boundary map inversion to find peaks



Boundaries generated using Canny method









### 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







F Global motion





stimuli





Whole



Part







Inconsistent

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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<sup>a,1</sup>, Hang Joon Jo<sup>b,1,2</sup>, Gregory L. Wallace<sup>a</sup>, Ziad S. Saad<sup>b</sup>, Robert W. Cox<sup>b</sup>, and Alex Martin<sup>a</sup>

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**

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#### Left-lateralized Effects (P<.005):



#### Right-lateralized Effects (P<.005):

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



(RH)

#### Lateralization Magnitude Predicts Cognitive Ability

#### **Lateralization Magnitude Predicts Cognitive Ability**



Examples:

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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,<sup>1</sup> W. Kyle Simmons,<sup>2</sup> Lydia A. Milbury,<sup>1</sup> Gregory L. Wallace,<sup>1</sup> Robert W. Cox<sup>3</sup> and Alex Martin<sup>1</sup>

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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





4.0



#### 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







*t-val* **TD - COS** (df = 43)



Group Differences Are Associated with Positive/Negative Symptoms in COS



Correlation with Positive Symptoms (SAPS) (covarying Age, Motion)



Correlation with Negative Symptoms (SANS) (covarying Age, Motion)



#### **Increased Resting Correlations in Primary Lateral Sclerosis (PLS)**

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

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



Controls



PLS



**PLS-Control** 



**Correlation with** 



#### **5** Patients with Left Hemisphere Lesions due to left CVA

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

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## **Example Patient**

### MPRAGE





# **Example Patient**

### MPRAGE

EPI



## Example Patient MPRAGE

EPI







## **Example Patient**

#### **MPRAGE**

#### **Tissue Masks**



## Lesion Reconstruc<mark>tion</mark>







## **Example Patient (After Data Cleaning)**

**R** Postcentral Gyrus Seed (r > +0.35):



**R** Intraparietal Sulcus Seed (r > +0.35):







## **Example Patient (After Data Cleaning)**



## **Example Patient (After Data Cleaning)**



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),



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

- 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)







log(# RH voxels > threshold)

threshold = .2 threshold = .3





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





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



## Summary

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- 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