



MRC Cognition
and Brain
Sciences Unit



UNIVERSITY OF
CAMBRIDGE

EEG/MEG 3:

Advanced Functional Connectivity Analysis

Olaf Hauk

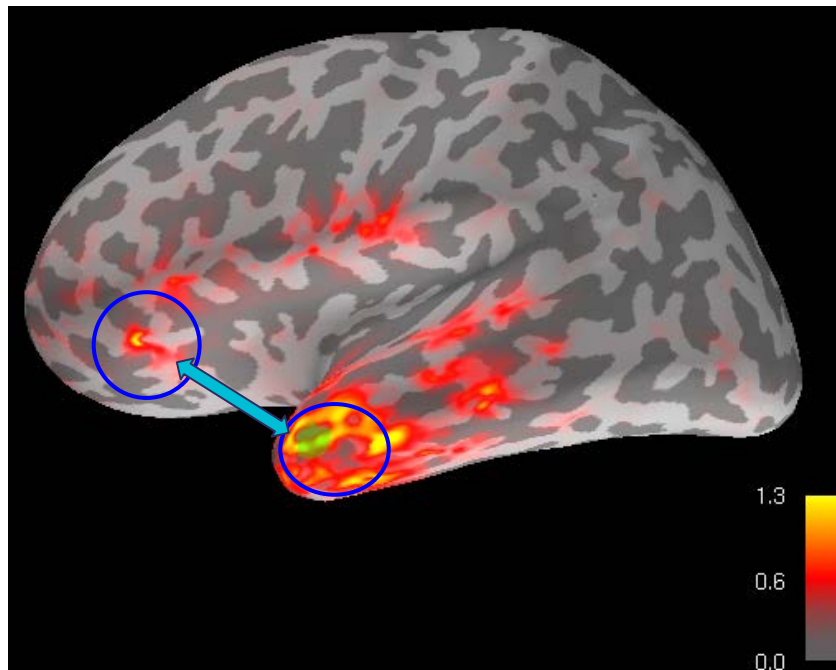
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COGNESTIC 2022

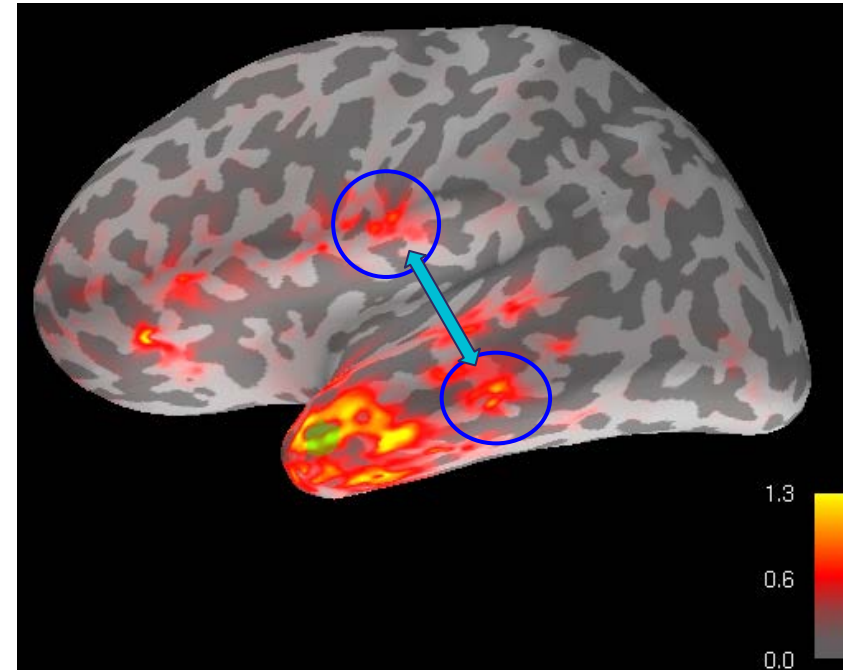
Spatial Resolution And Leakage Can Confound Connectivity Measures

Field Spread / Point Spread

Connectivity between two regions may reflect cross-talk from one of the regions



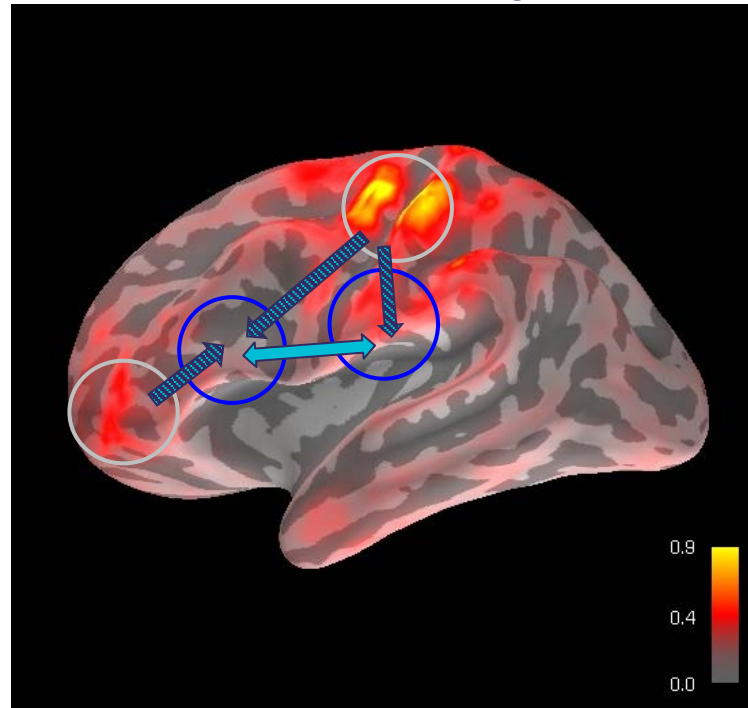
Connectivity between two regions may reflect cross-talk from a third region



Some connectivity measures can rule out “zero-lag” connectivity
(but they are then also insensitive to real zero-lag connectivity)

Field Spread / Point Spread

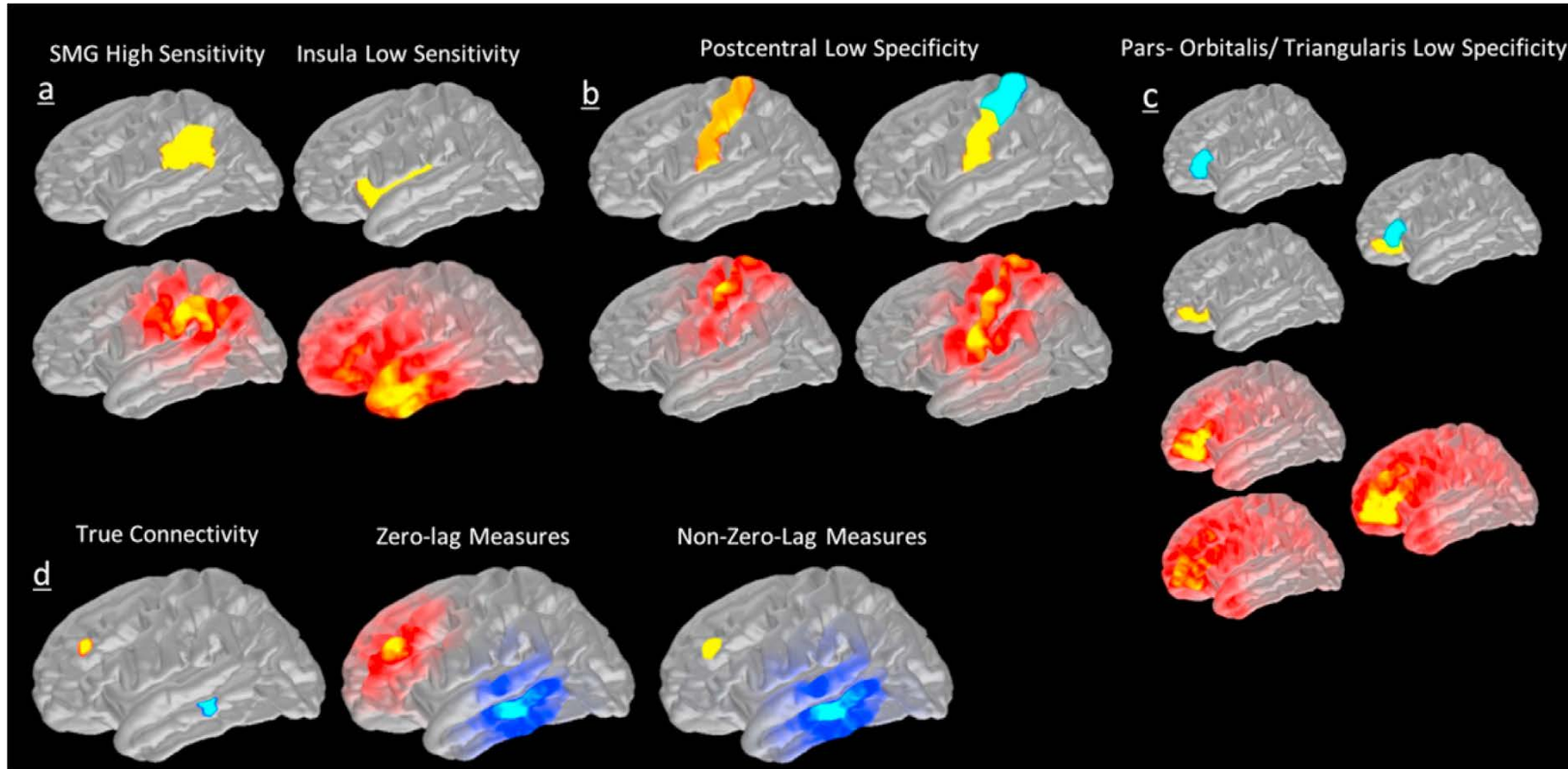
Connectivity between two regions
may reflect cross-talk from
several other regions



This is bad, and there is not much you can do –
except getting your model right in the first place, or use whole-brain
analysis.

Leakage Can Produce Spurious Connectivity

(also at zero-lag)



Farahibozorg, Henson, Hauk, NI 2018, <https://pubmed.ncbi.nlm.nih.gov/28893608/>

See also:

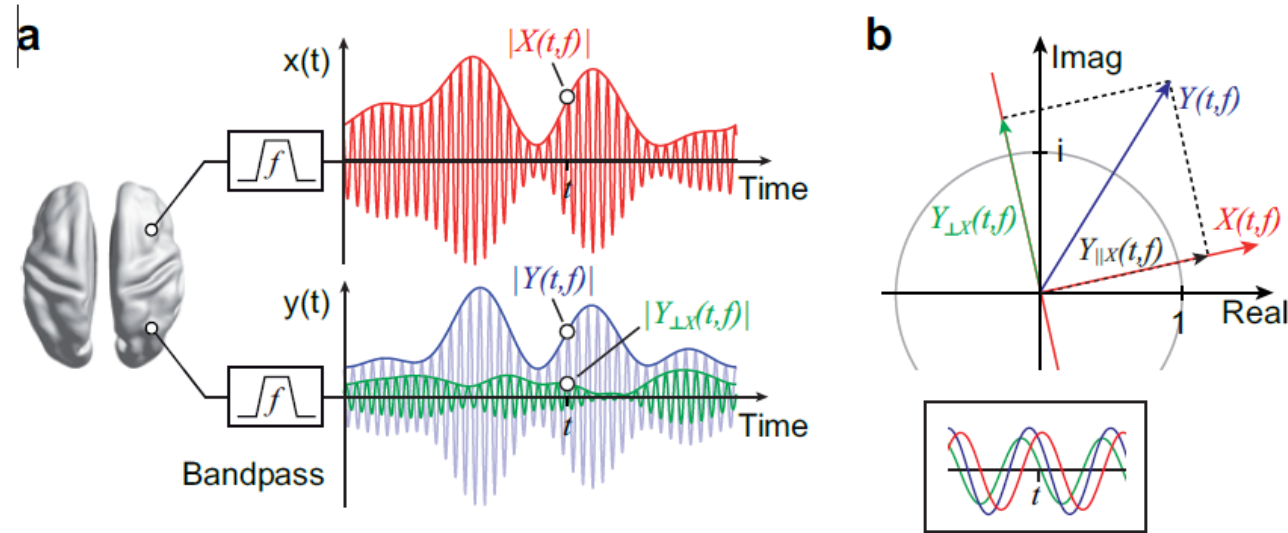
Palva et al., NI 2018, <https://pubmed.ncbi.nlm.nih.gov/29477441/>

Colclough et al. NI 2015, <https://pubmed.ncbi.nlm.nih.gov/25862259/>

One Possibility: Remove Zero-Lag Connectivity

Orthogonalisation of time courses, Partial regression

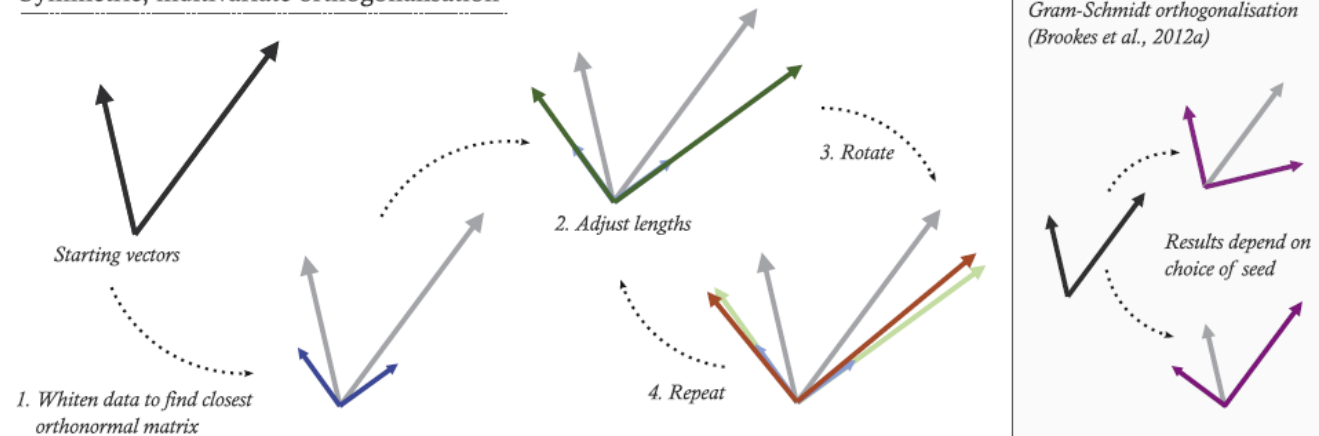
Bivariate:



Hipp et al., Nat Nsc 2012, <https://www.nature.com/articles/nn.3101>

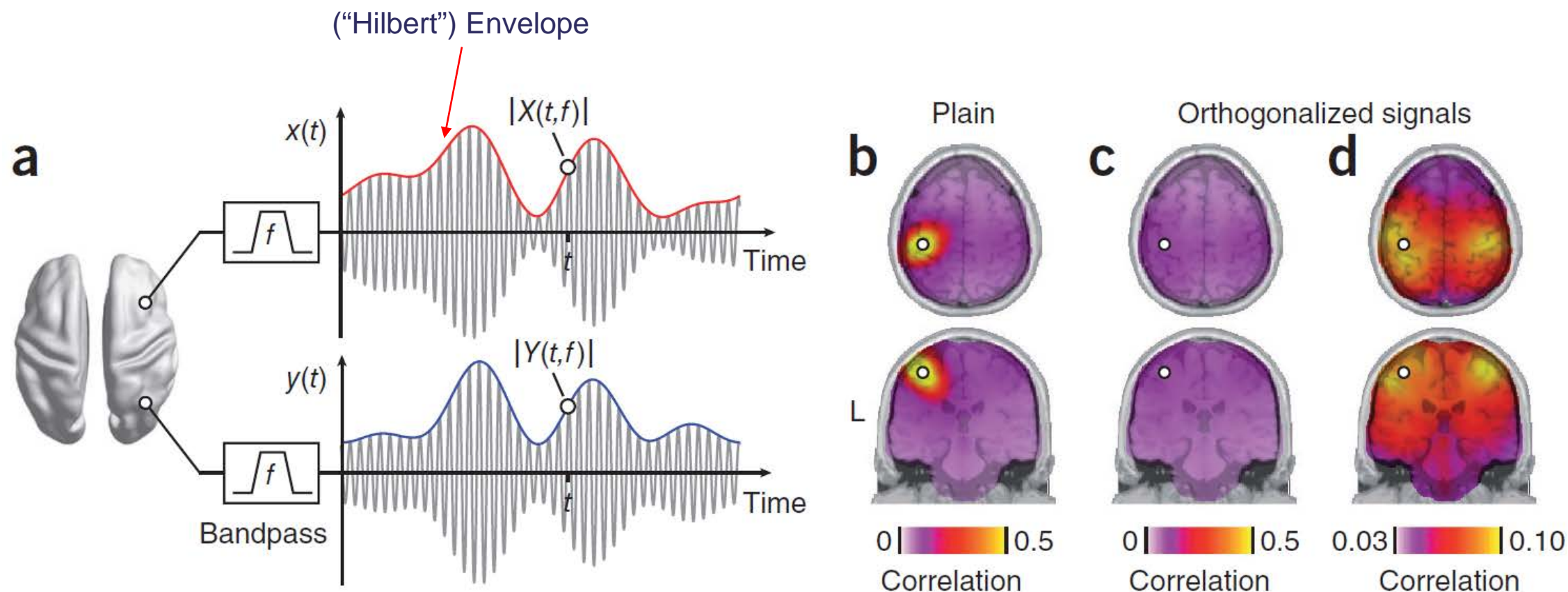
Multivariate:

Symmetric, multivariate orthogonalisation

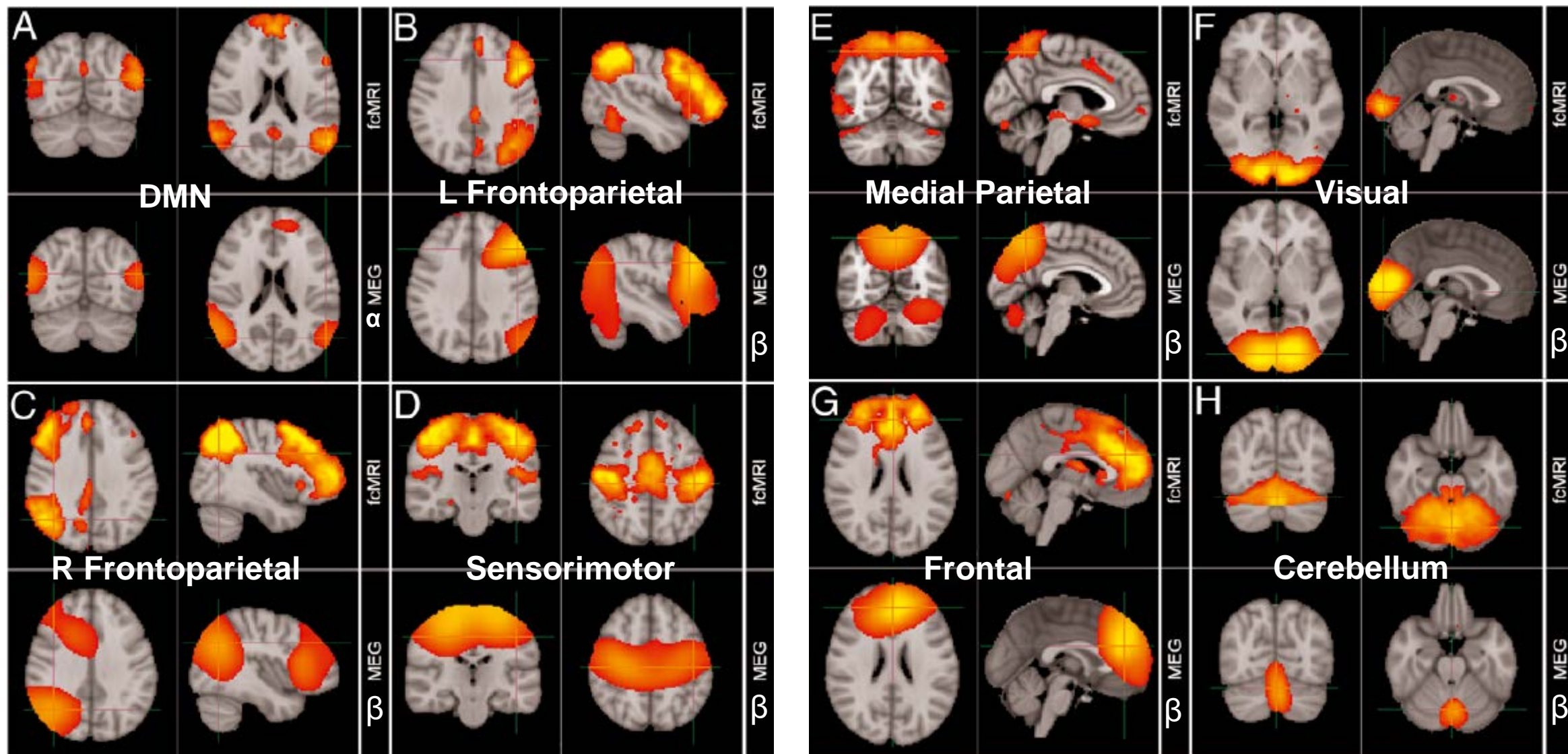


Colclough et al., NI 2015, <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC4528074/>

Functional Connectivity of Resting State Activity



Functional Connectivity of Resting State Activity



One Possibility: Remove “Zero-Lag” Connectivity

Imaginary Part of Coherency

In spectral connectivity measures like Coherence, only use the imaginary part of the signal, which is unaffected by zero-lag connectivity (phase differences of zero are only represented in the real part).

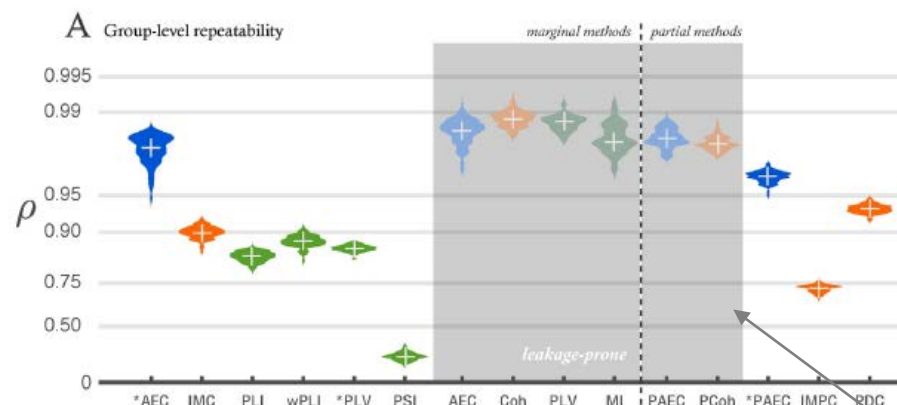
Ewald et al., NI 2012, <https://pubmed.ncbi.nlm.nih.gov/22178298/>

Pascual-Marqui, arXiv 2007a and 2007b, <https://arxiv.org/abs/0706.1776>, <https://arxiv.org/abs/0711.1455>

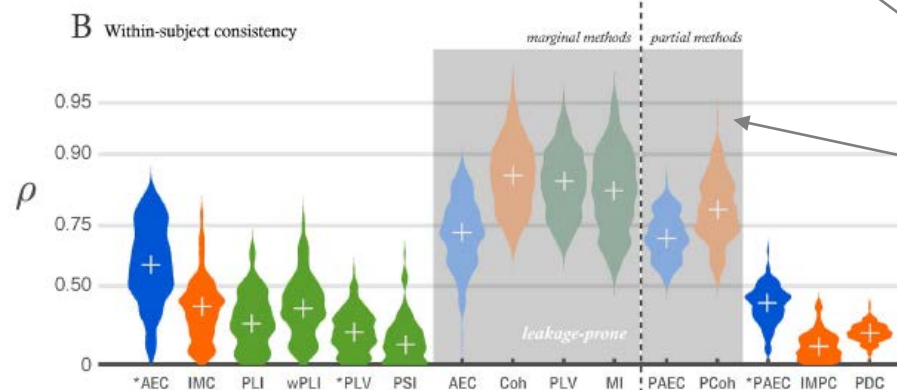
Note: “Non-zero-lag methods” may also ignore true zero-lag connectivity, e.g. for bilateral sources – one may throw out the child with the bath water.

Leakage and Reliability of Functional Connectivity Methods

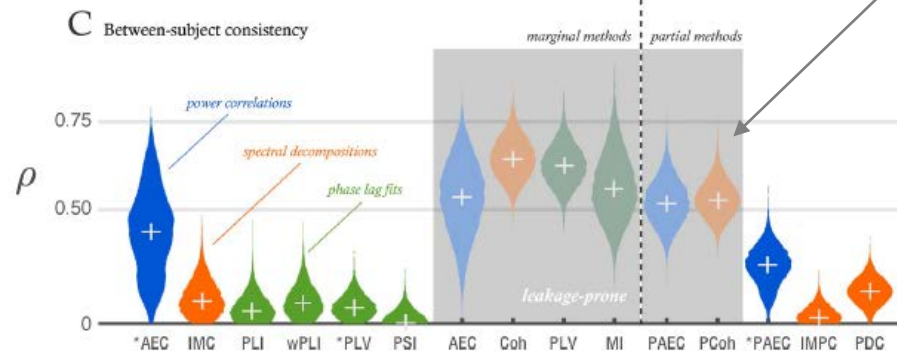
Group-level repeatability



Within-subject consistency



Between-subject consistency

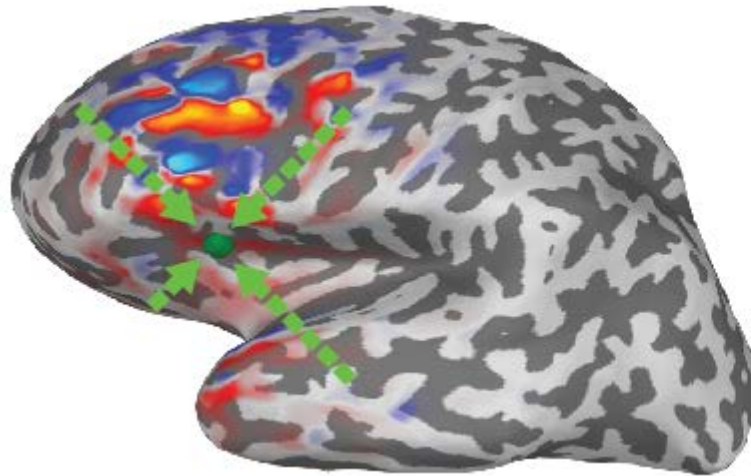


leakage-prone

Spatial Resolution / Leakage:

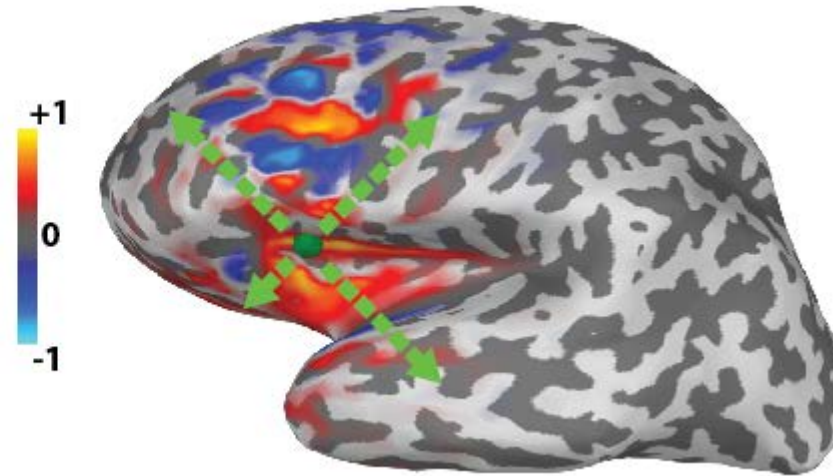
Point-Spread and Cross-Talk

Cross-Talk Function
(CTF)



How other sources may affect the estimate for this source

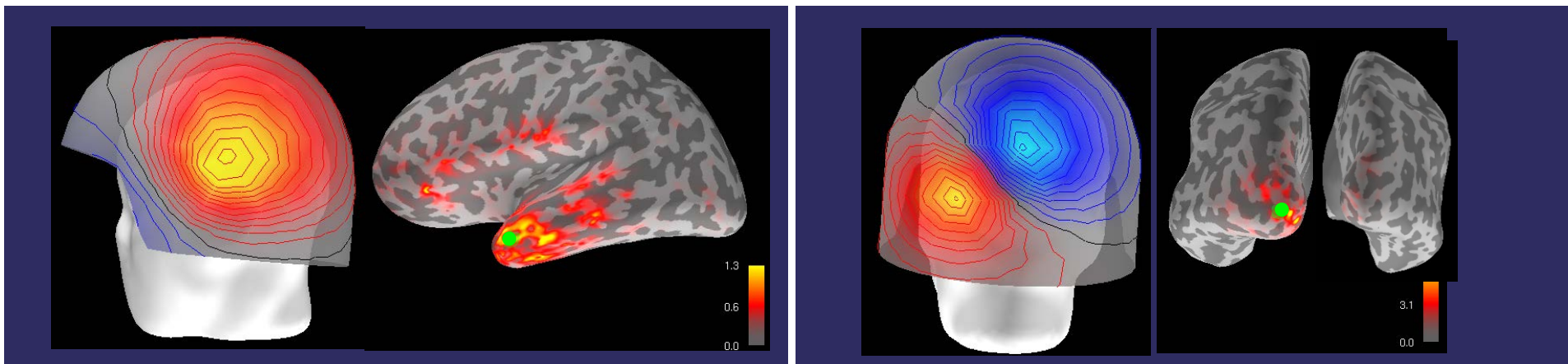
Point-Spread Function
(PSF)



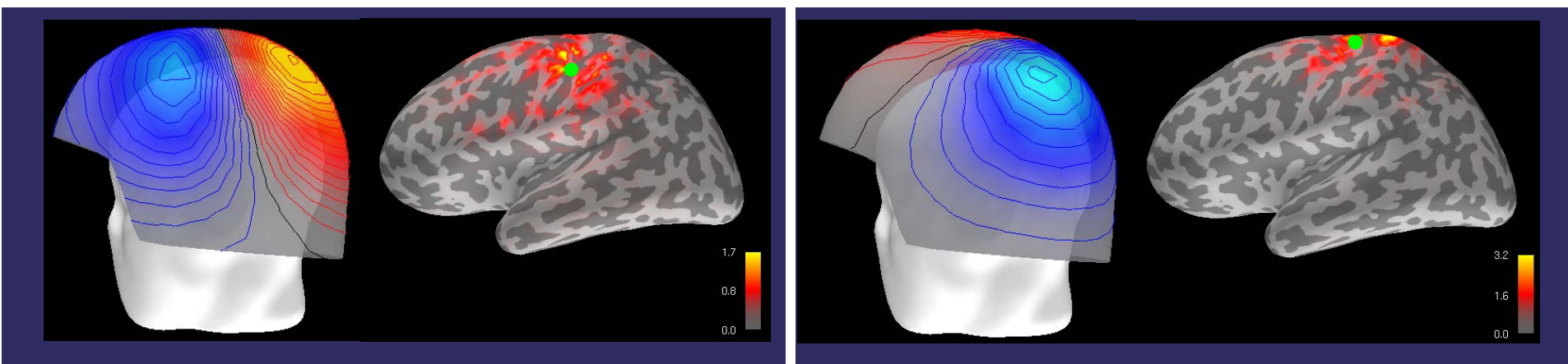
How this source affects estimates for other sources

PSFs and CTFs for Some ROIs

For MNE, PSFs and CTFs turn out to be the same

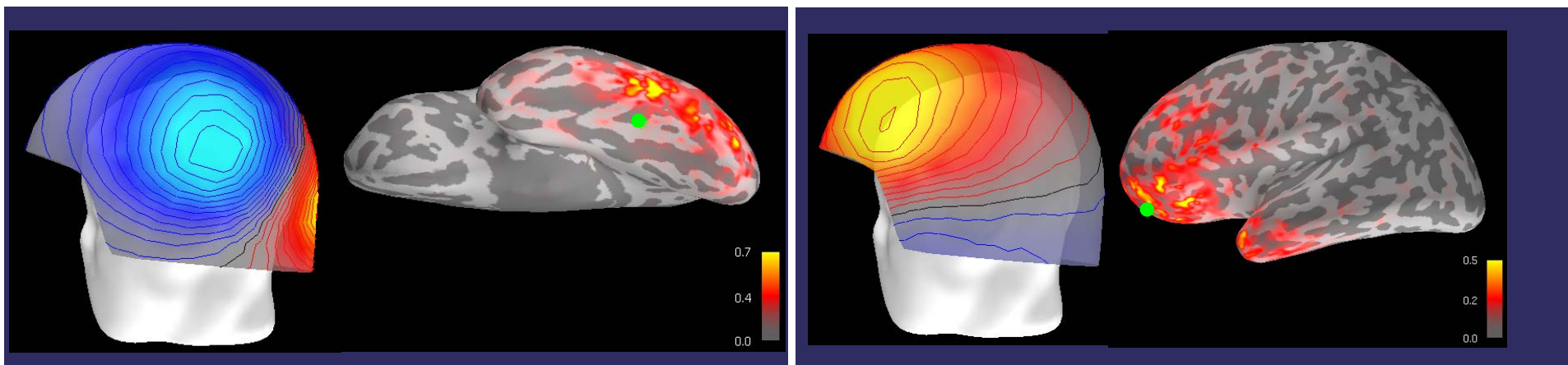


Good

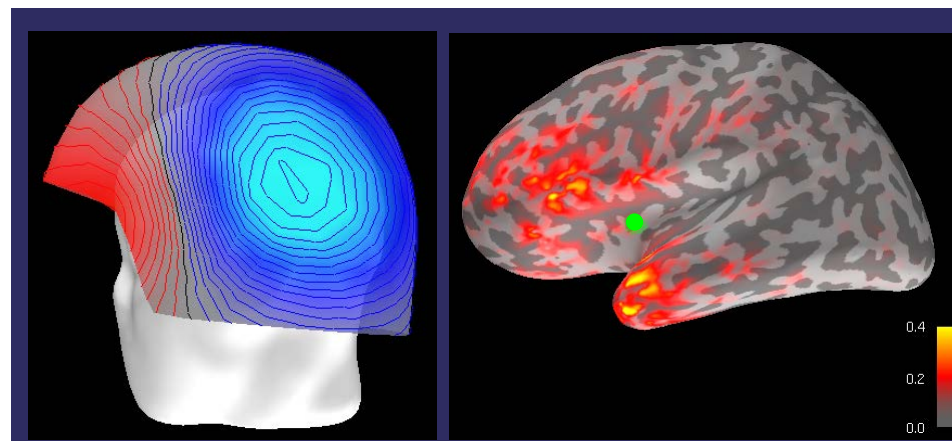


PSFs and CTFs for Some ROIs

For MNE, PSFs and CTFs turn out to be the same

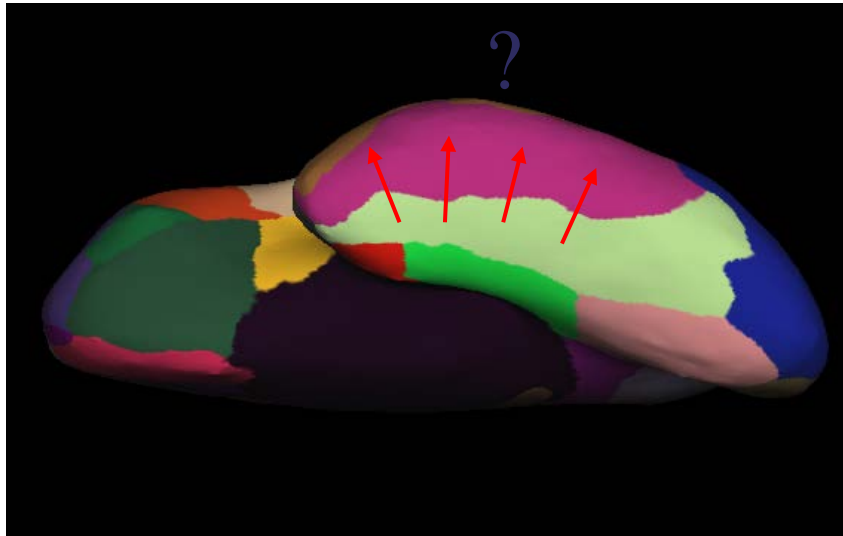


Less good

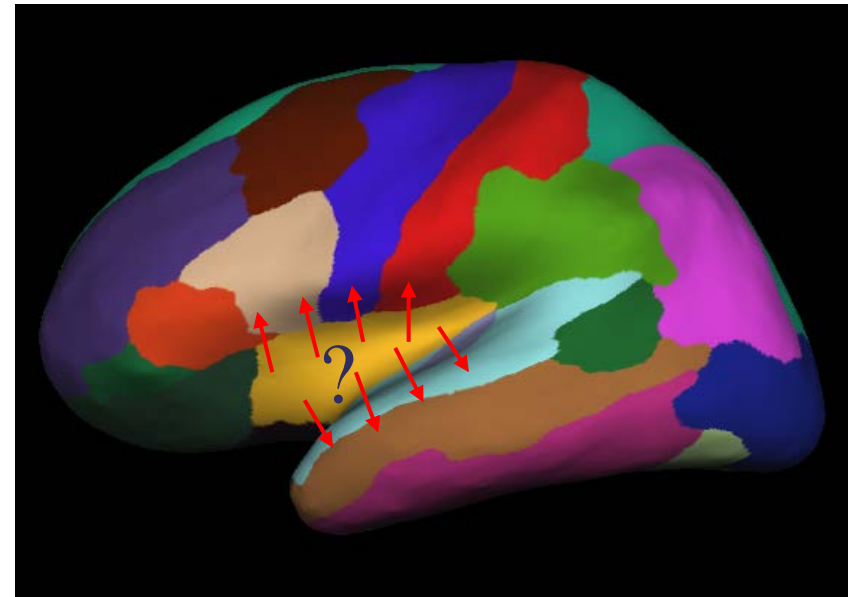


Localisation Bias Has Consequences for ROI analysis

PSFs/CTFs Can Tell You How It Looks Like

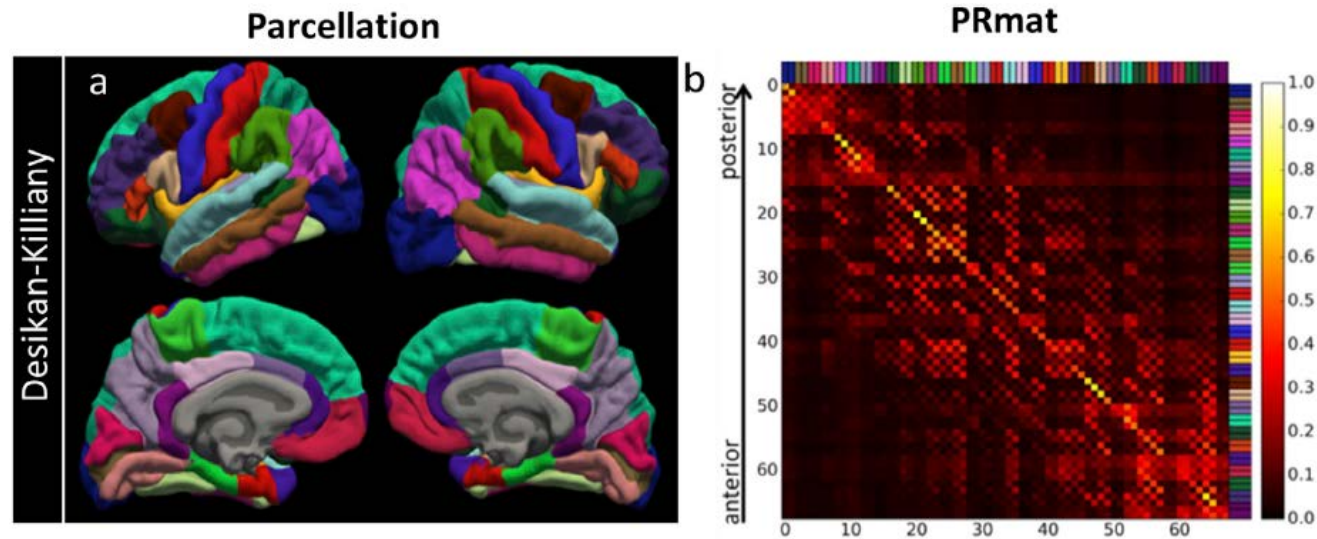


Desikan-Killiany Atlas parcellation

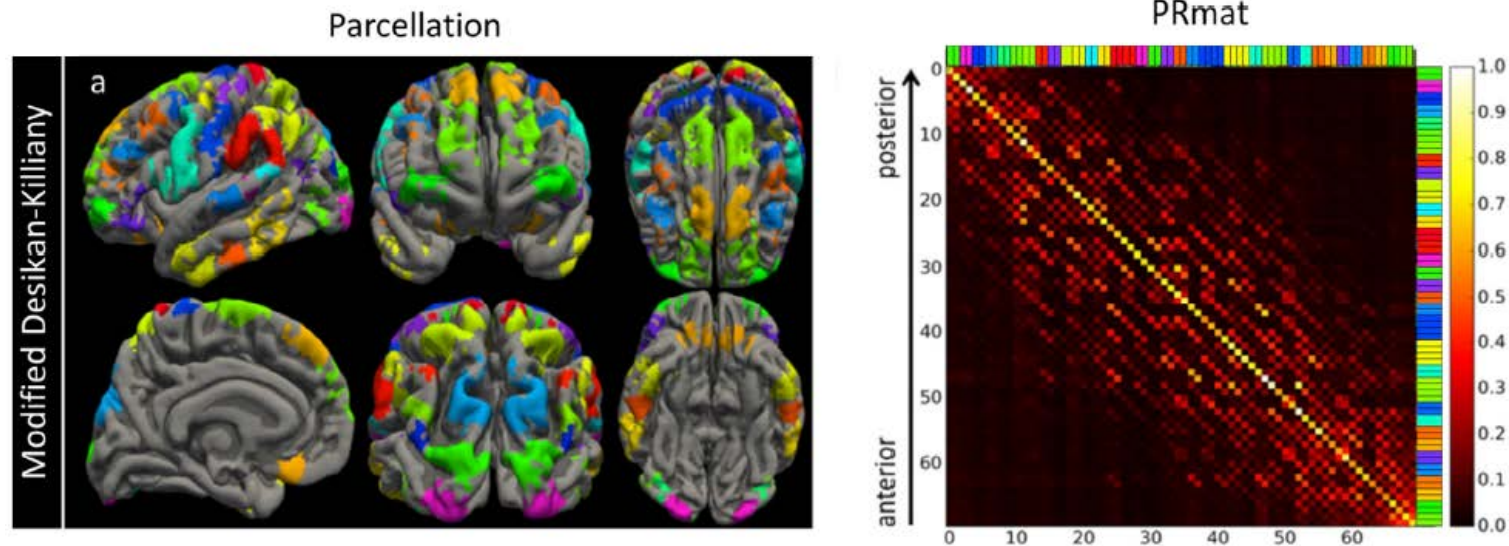


Adaptive cortical parcellation based on resolution matrix

Original Parcellation



Modified Parcellation



And Beyond...

Most of the previously introduced measures are spectral measures, i.e. they are computed for specific frequencies (or frequency bands).

They rely on the assumption that brain signals can meaningfully be decomposed into “oscillations” or “frequency bands”.

This is a big assumption, and may not be the case for all modalities, stimuli, tasks etc., or may not even be true in general.

Therefore...

Non-Spectral and Effective Connectivity

Granger Causality: Is one time series useful to predict another?
 $x(t)$ Granger-causes $y(t)$ if past values of $x(t)$ add information to past values of $y(t)$ for predicting future values of $y(t)$.

http://www.scholarpedia.org/article/Granger_causality

Multivariate Granger Toolbox: <http://www.sussex.ac.uk/sackler/mvgc/>

<http://journal.frontiersin.org/article/10.3389/fnsys.2015.00175/full>

Structural Equation Modelling (SEM):

Models covariance structure of brain activation across brain regions (e.g. “path analysis”).

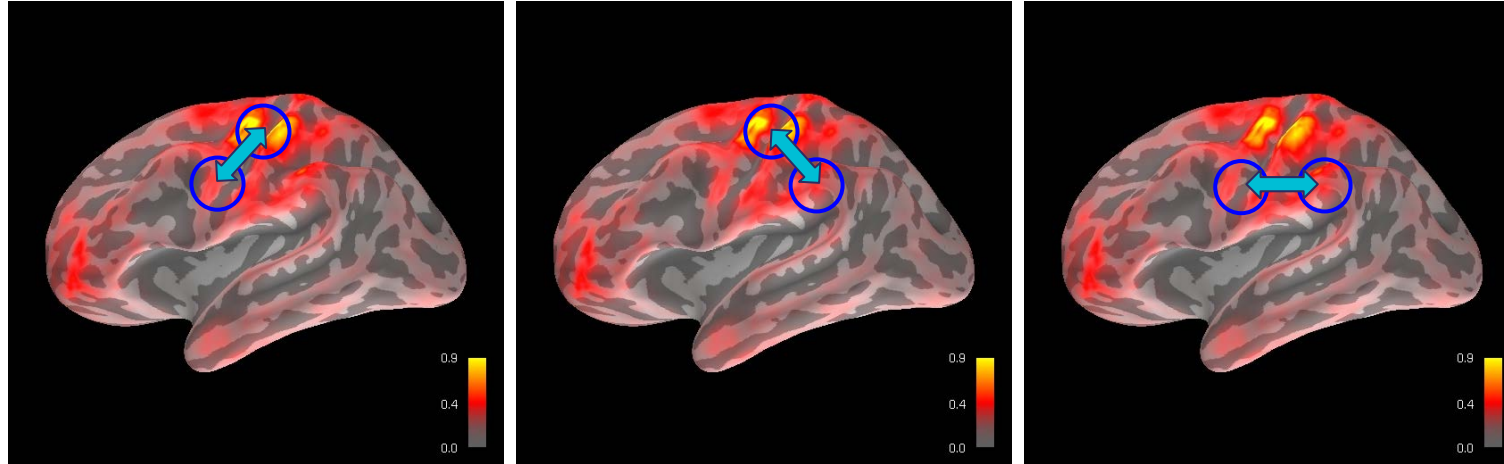
Dynamic Causal Modelling (DCM):

Models brain dynamics across regions as differential equations, in combination with Bayesian parameter/model estimation.

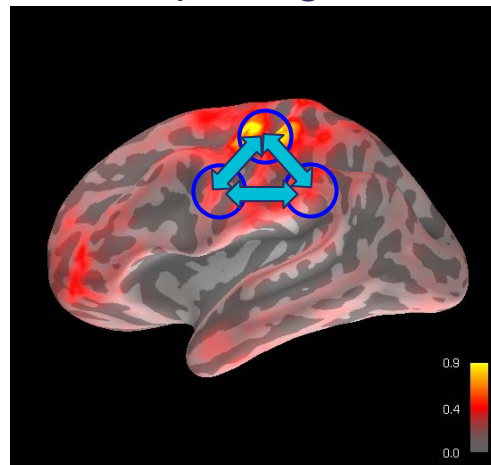
http://www.scholarpedia.org/article/Dynamic_causal_modeling

Bivariate vs Multivariate Connectivity

Bivariate measures test one pair or regions at a time:

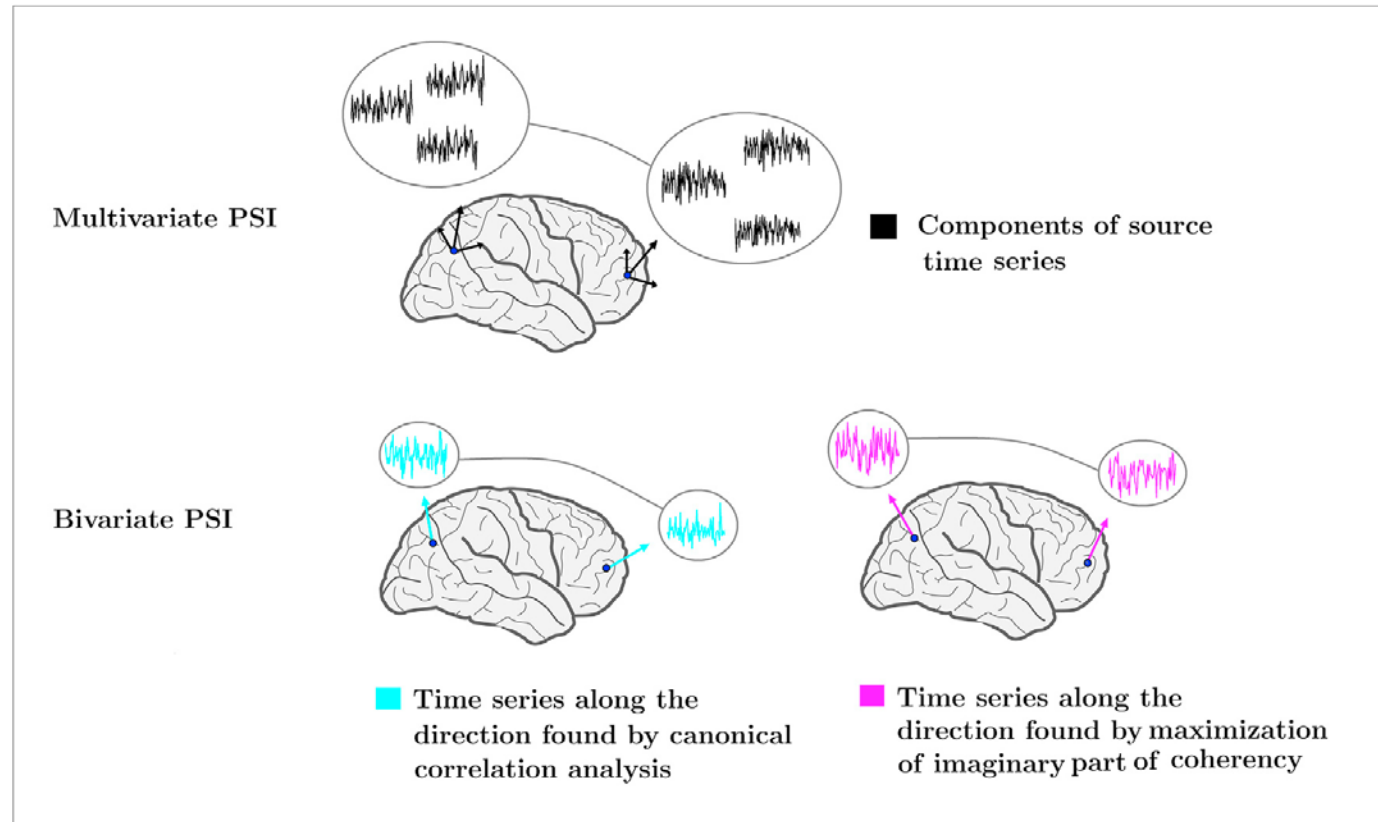


Multivariate measures test multiple regions simultaneously:



Multi-Variate and Multi-Dimensional Connectivity

Currently, most connectivity methods use one time course per ROI. However, brain activity is multivariate, and there is potentially a lot of information lost by collapsing across vertices or voxels. “Multi-dimensional” methods are now emerging.



Basti et al., NI 2018, <https://www.sciencedirect.com/science/article/pii/S1053811918301897>

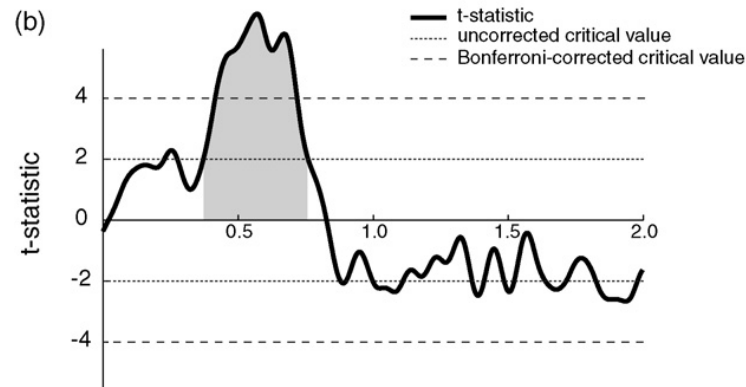
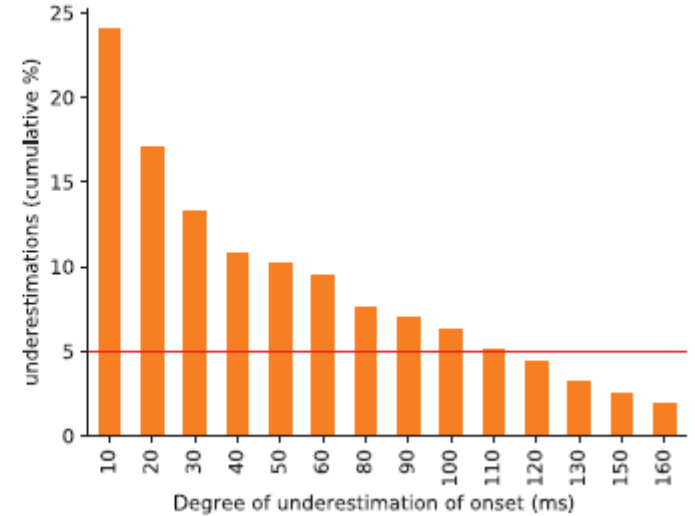
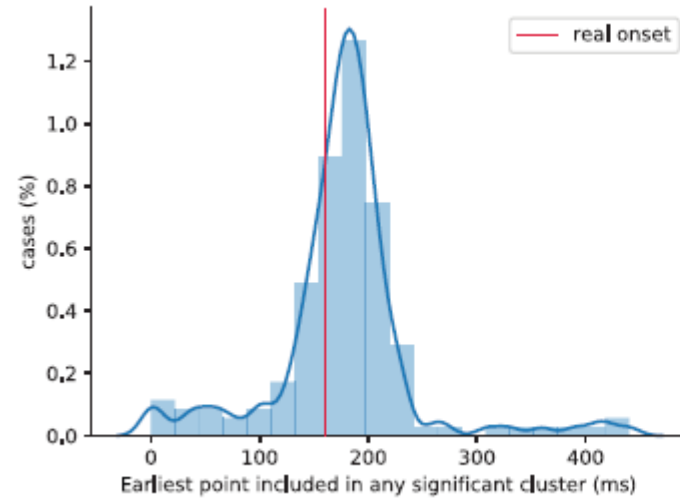
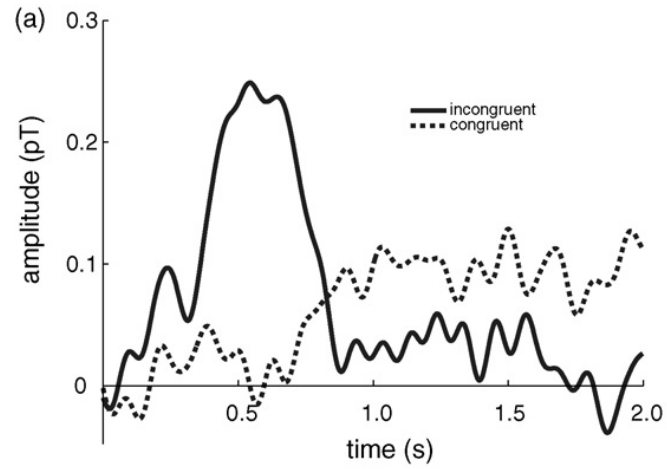
Also:

Basti/Nili et al., NI 2020, <https://www.sciencedirect.com/science/article/pii/S1053811920306650>

Anzellotti & Coutanche, T Cogn Sci 2018, <https://pubmed.ncbi.nlm.nih.gov/29305206/>

Basti et al., PLoS 2019, <https://journals.plos.org/plosone/article/comments?id=10.1371/journal.pone.0223660>

Statistics – Cluster-Based Permutation Tests



Cluster-based permutation tests of MEG/EEG data do not establish significance of effect latency or location

Sassenhagen & Draschkow, Psychophysiol 2018,
<https://pubmed.ncbi.nlm.nih.gov/28893608/>



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Thank you