

Covariance

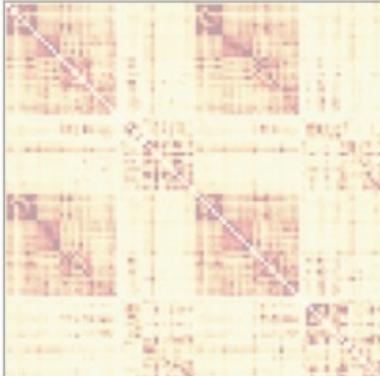
Precision

Spearman's ρ

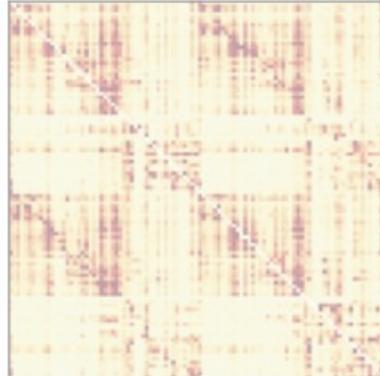
Pairwise distance
(Euclidean)

Benchmarking methods for mapping functional connectivity in the brain

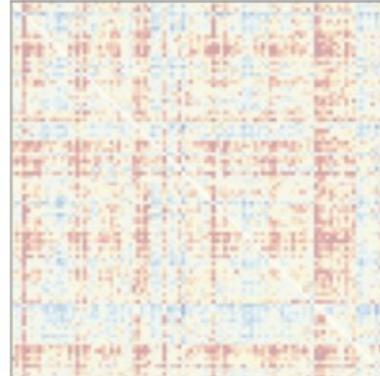
Mutual information
(Gaussian)



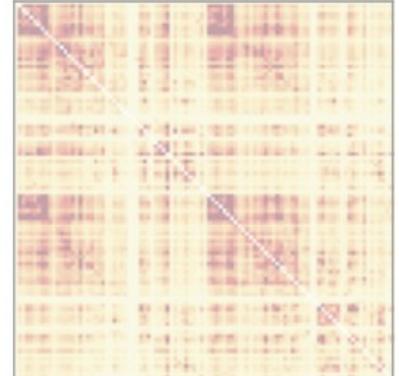
Granger causality
(Gaussian, $k = 1$, $kt = 1$, $l = 1$, $lt = 1$)
(mean, $fs = 1$, $f = [0.25, 0.5]$)



Coherent phase
(max, $fs = 1$, $f = [0.25, 0.5]$)

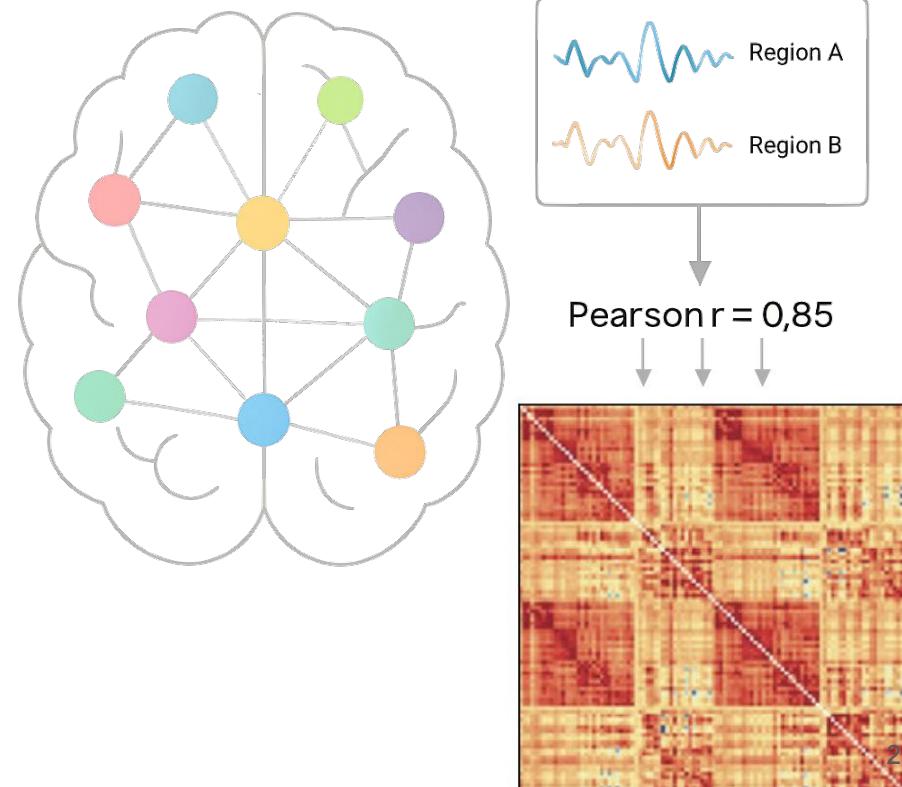


Phase lag value
(max, $fs = 1$, $f = [0.25, 0.5]$)



What is functional connectivity

- Brain connectivity describes how brain regions are linked and communicate.
- Functional connectivity (FC) describes it using pairwise statistical tests on the signals from different brain regions.
- FC is typically estimated as Pearson correlation coefficients (PCC), which reflect the covariance between signals.

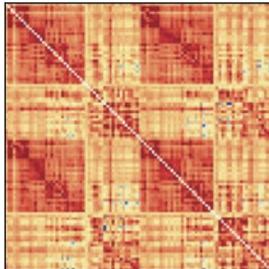


What this paper did

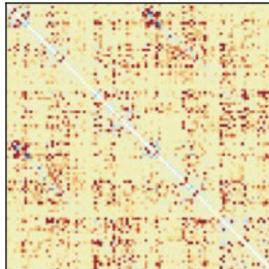
There is more to pairwise statistics than PCC, different tests can highlight different things.

This paper compared lots of these tests on a couple of connectivity benchmarks.

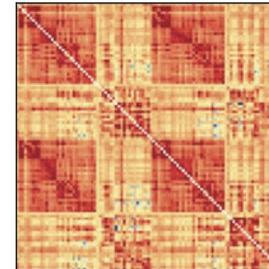
Covariance
(empirical covariance)



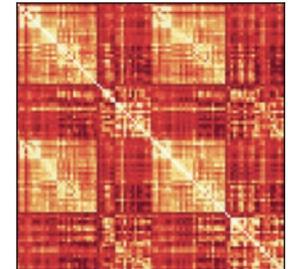
Precision
(empirical covariance)



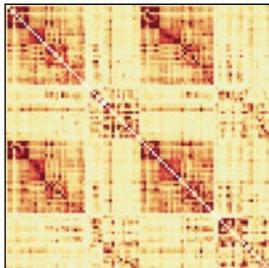
Spearman's ρ



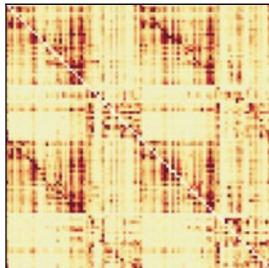
Pairwise distance
(Euclidean)



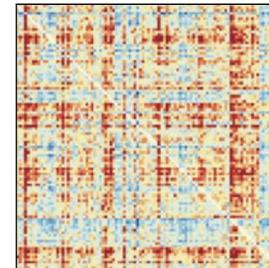
Mutual information
(Gaussian)



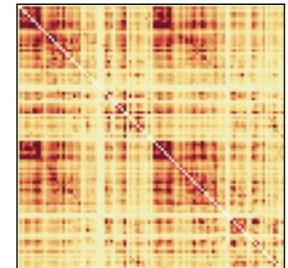
Granger causality
(Gaussian, $k = 1$, $kt = 1$, $l = 1$, $lt = 1$) (mean, $fs = 1$, $f = [0.25, 0.5]$)



Coherent phase



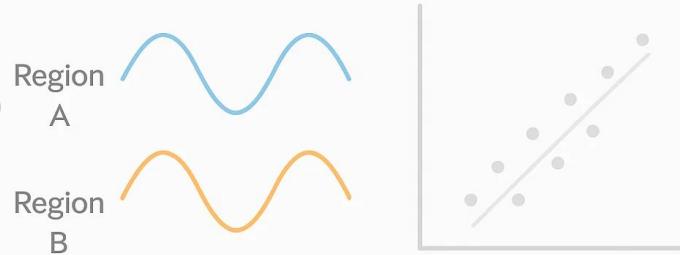
Phase lag value
(max, $fs = 1$, $f = [0.25, 0.5]$)



Families of pairwise statistics

Covariance
(‘Basic’ in paper)

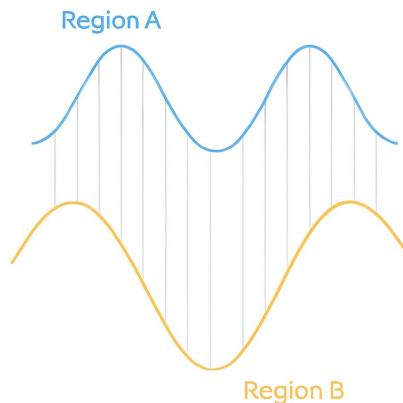
*Pearson
Precision*



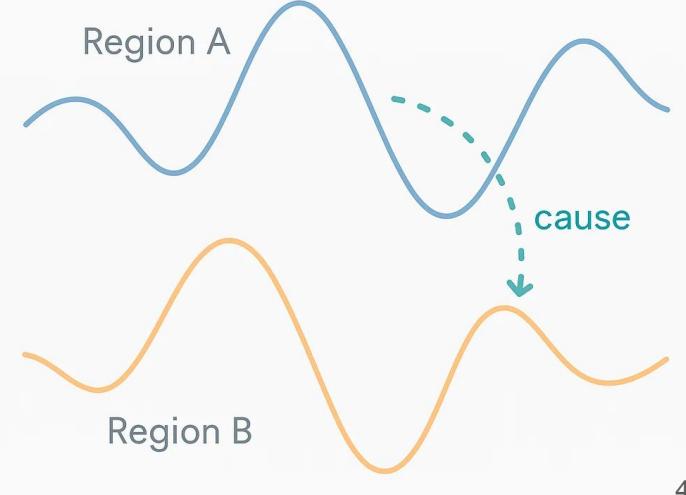
Distance

Pairwise distance

Dynamic Time Warping



Causal
Additive Noise Model
Conditional-Distribution Similarity

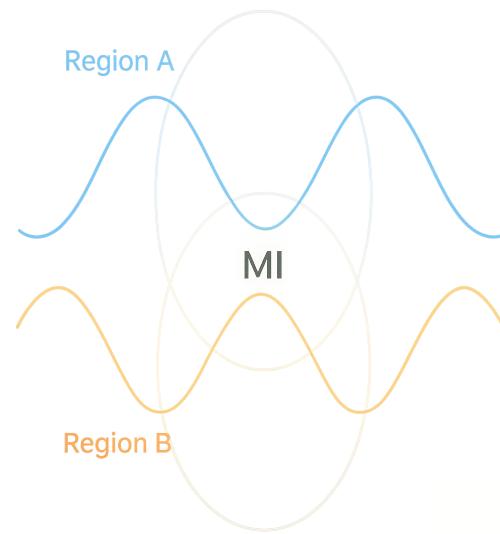


Families of pairwise statistics

Info theory

Mutual Information

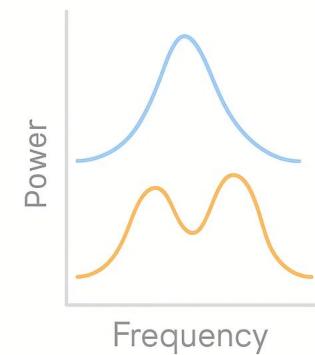
Joint entropy

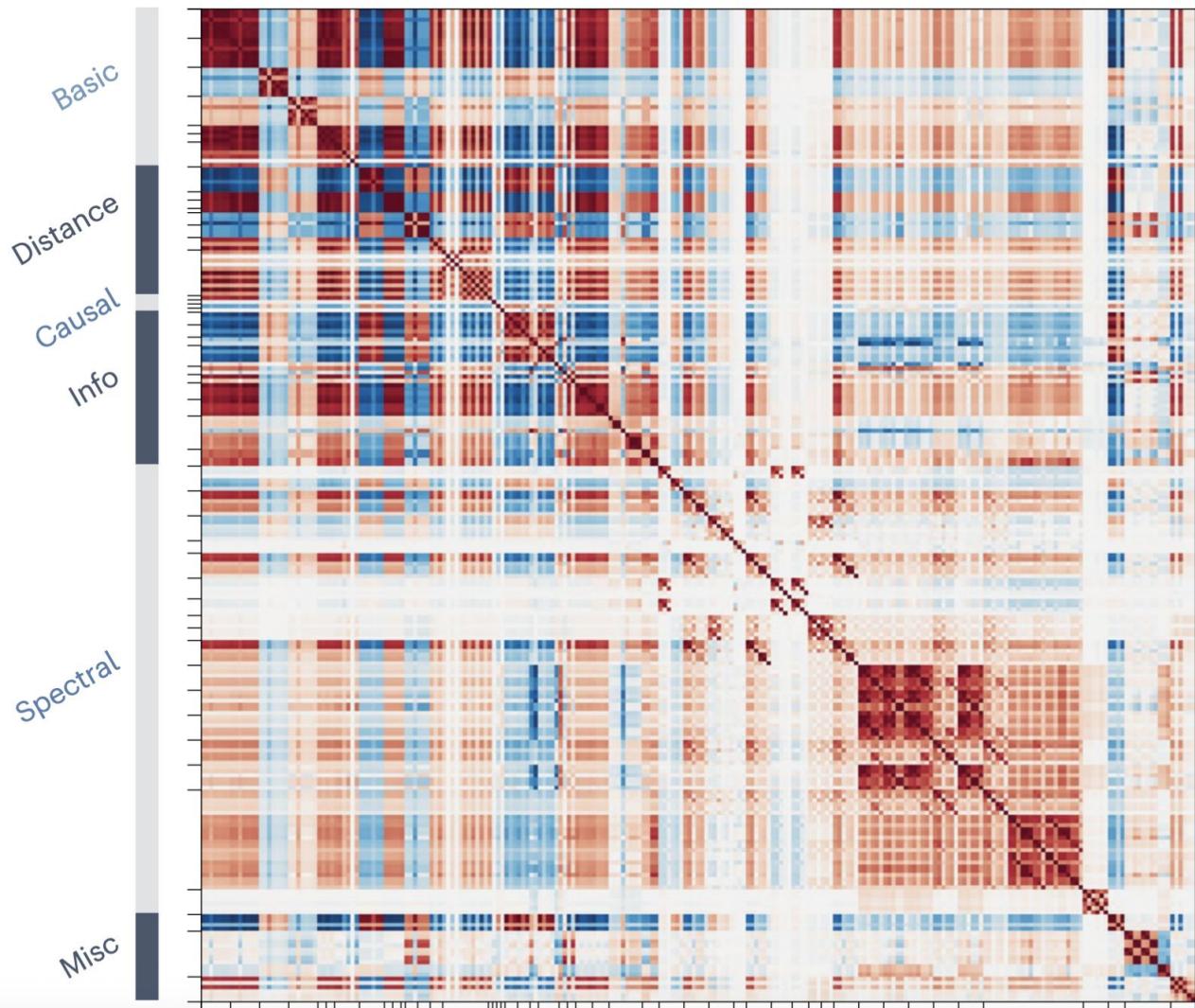


Spectral

Coherence

Phase Locking Value





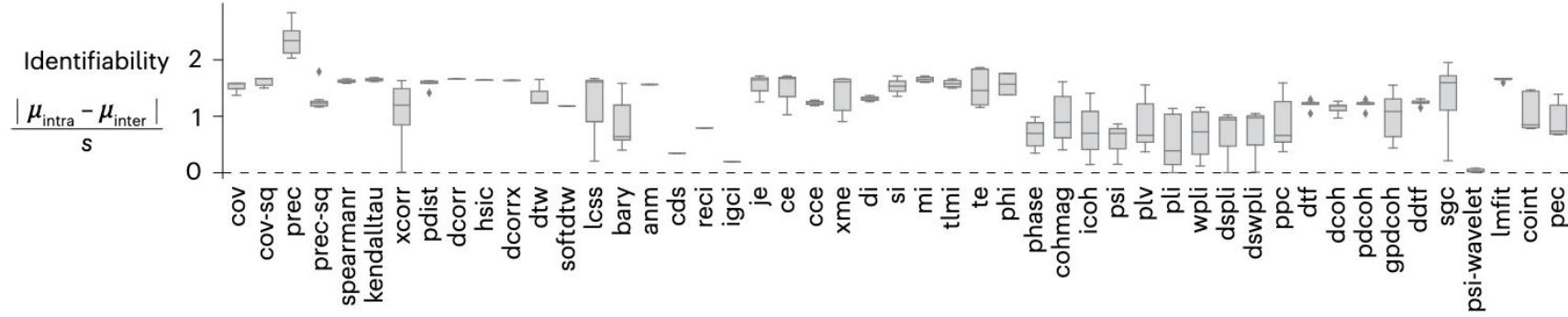
Average similarity between all pairs of 239 pairwise statistics.

Spearman ρ

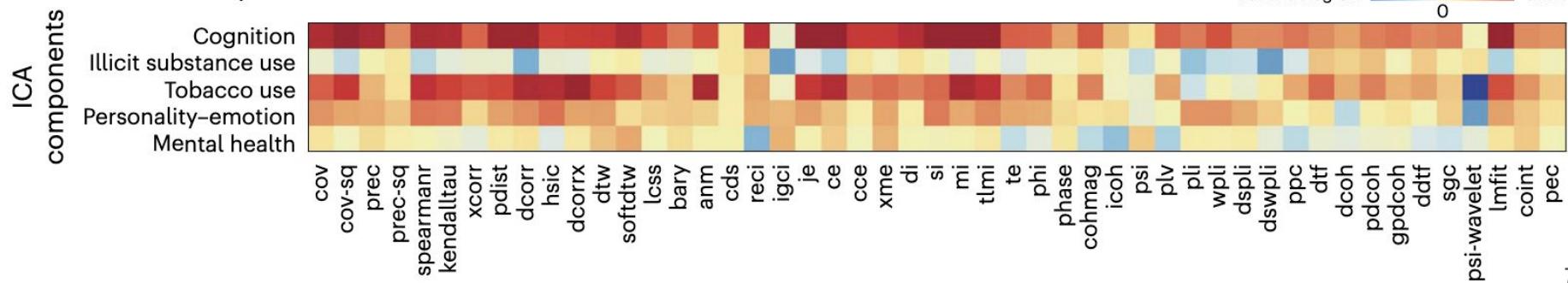
-1 1

Subject identifiability and behavior prediction

a Fingerprinting



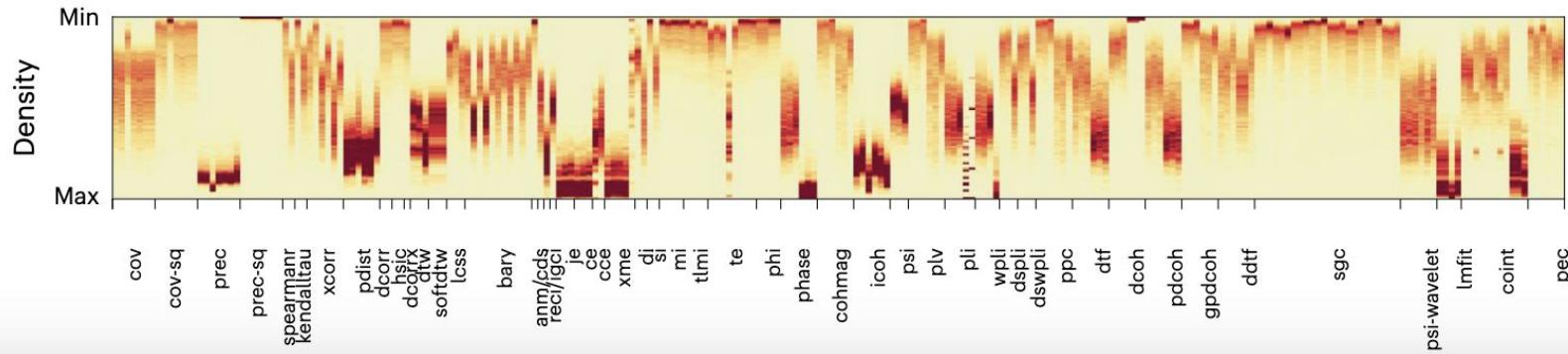
b Brain-behavior prediction



Connectivity benchmarks

Connectivity strength variation

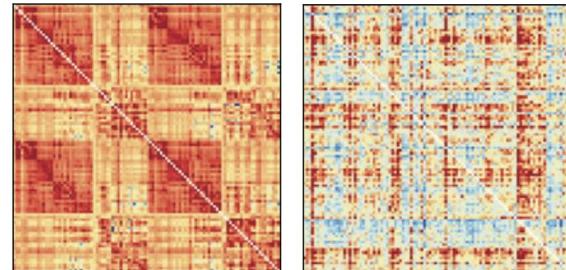
a Value distribution



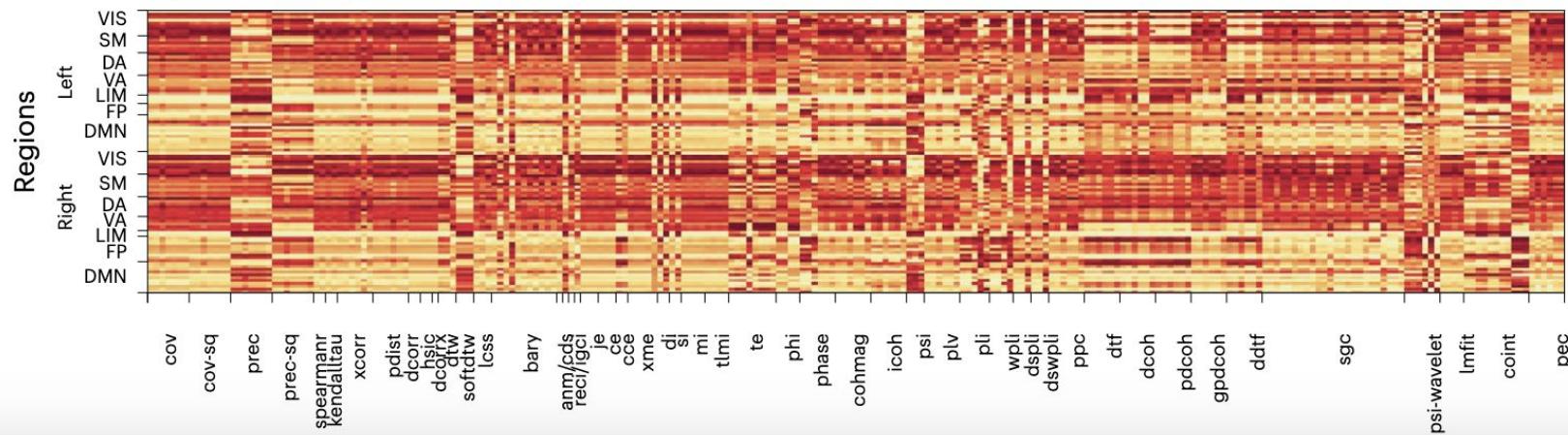
Connectivity benchmarks

Hubness. Good hub region has many connections to other regions.

Degrees of hubness



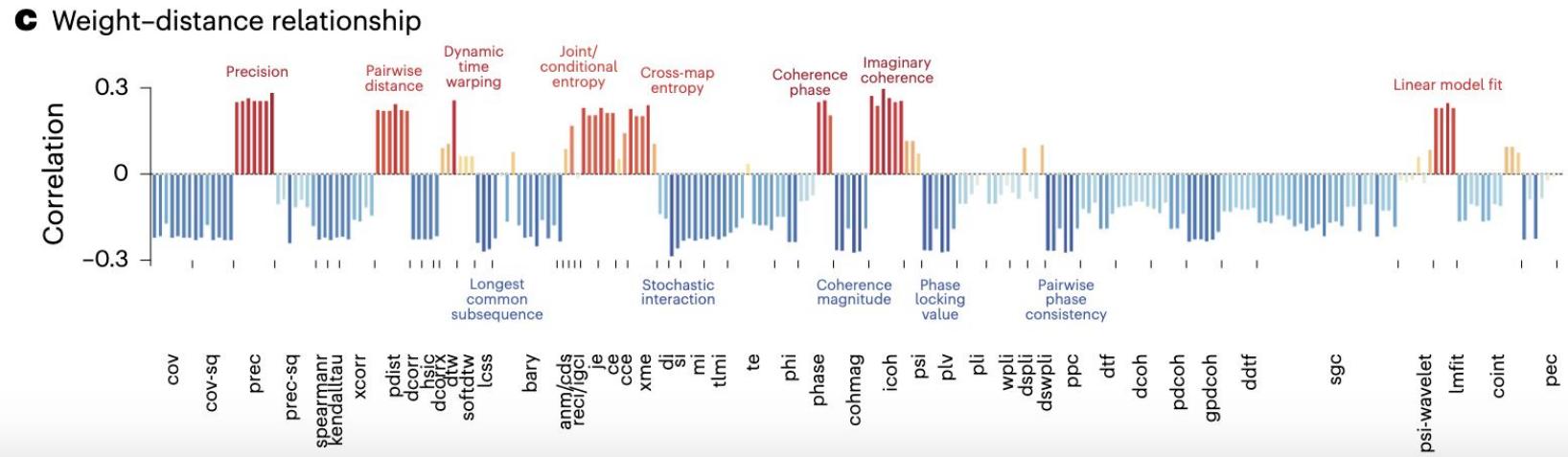
b Hubness (weighted degree)



Connectivity benchmarks

Weight-distance relationship.

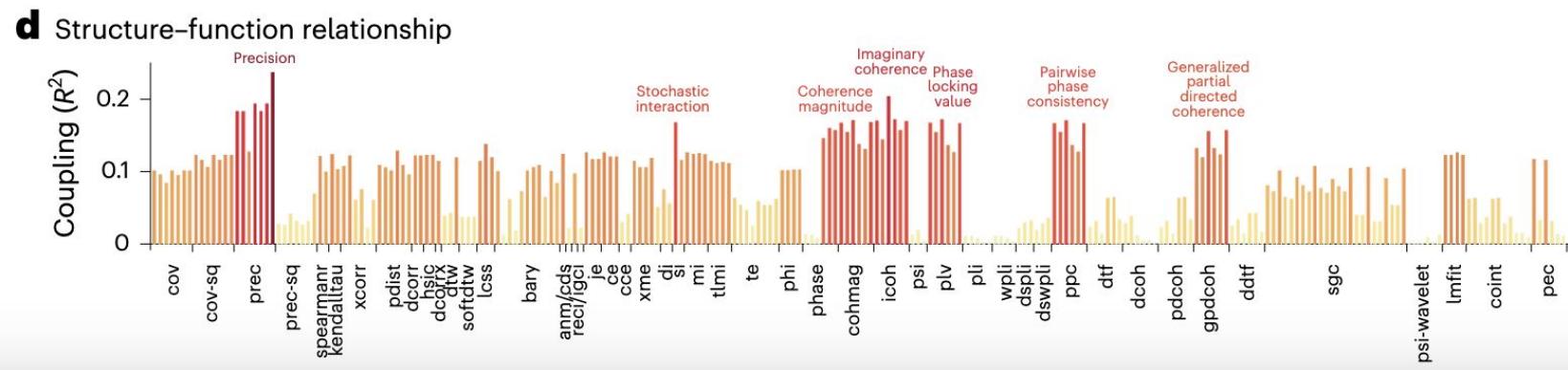
Distant brain regions tend to have weaker connections.



Connectivity benchmarks

Structure-function relationship.

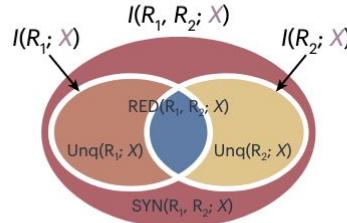
FC should be similar to the anatomy-based structural connectivity derived from diffusion MRI.



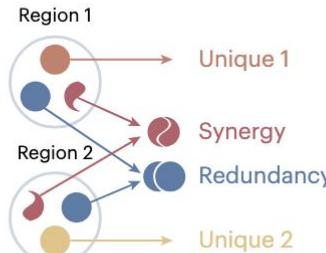
Integrated Information Decomposition

a Decomposing information flow

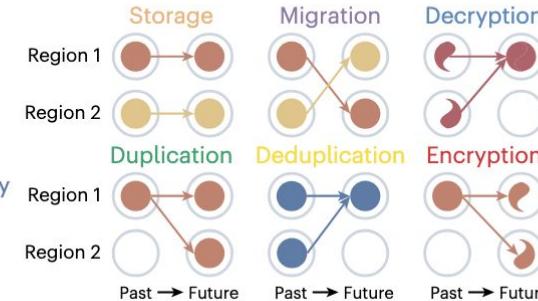
Picturing multivariate information



Defining synergy and redundancy



Categorizing information flow patterns

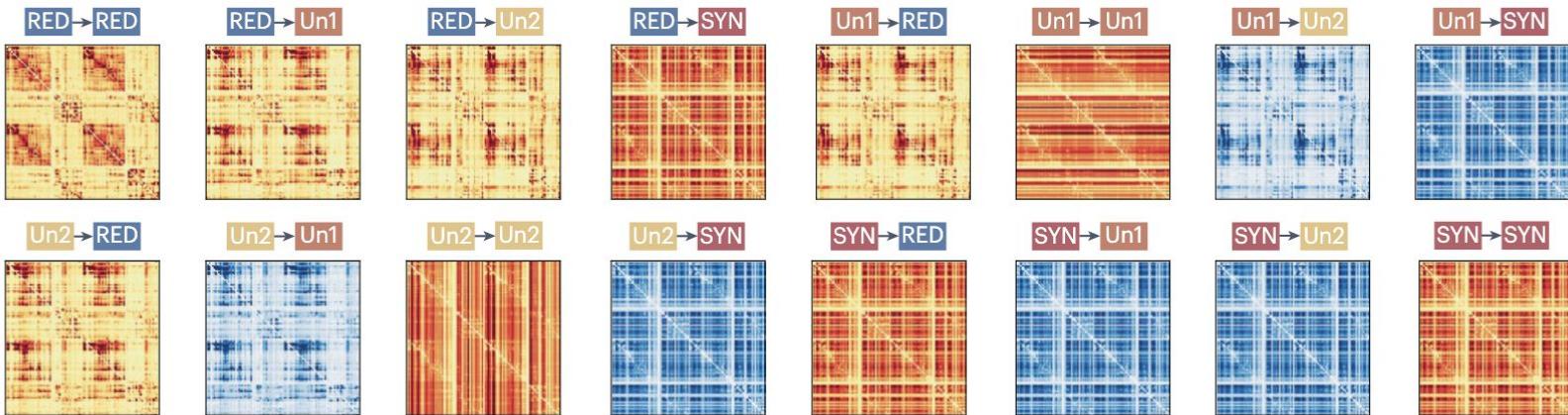


Formalizing pairwise transitions

		Future				
		SYN	Un1	Un2	RED	
		SYN	Stor	Decryp	Decryp	Decryp
		Un1	Encryp	Stor	Migr	Dupl
		Un2	Encryp	Migr	Stor	Dupl
		RED	Encryp	Dedup	Dedup	Stor

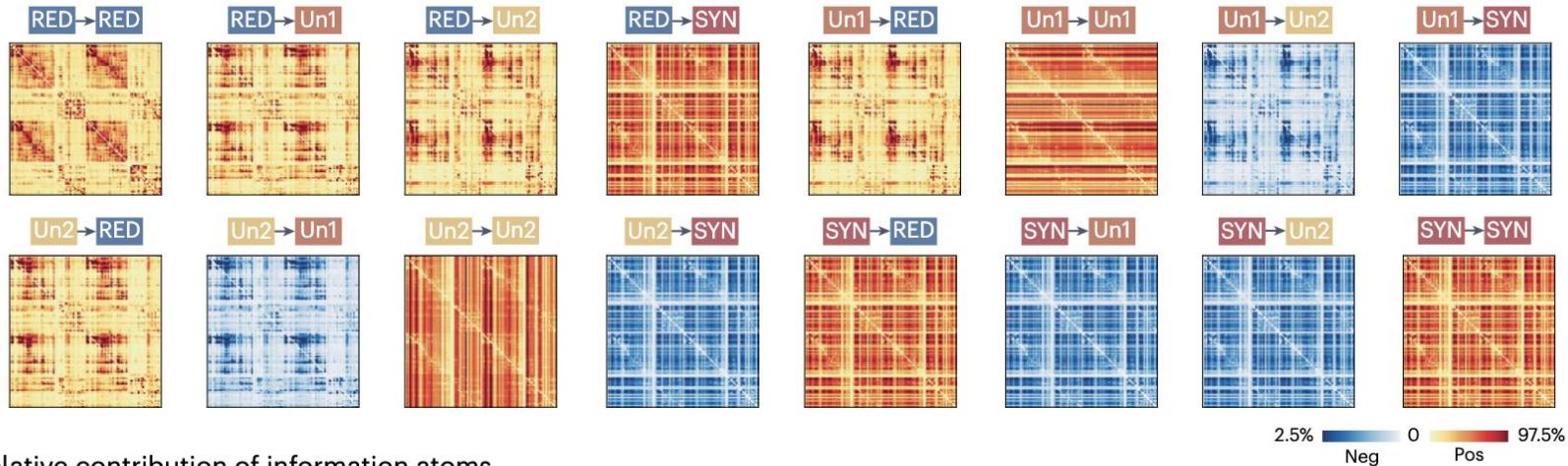
On the left, the word "Past" is written vertically above the first two rows of the table.

b Information-dynamic atoms

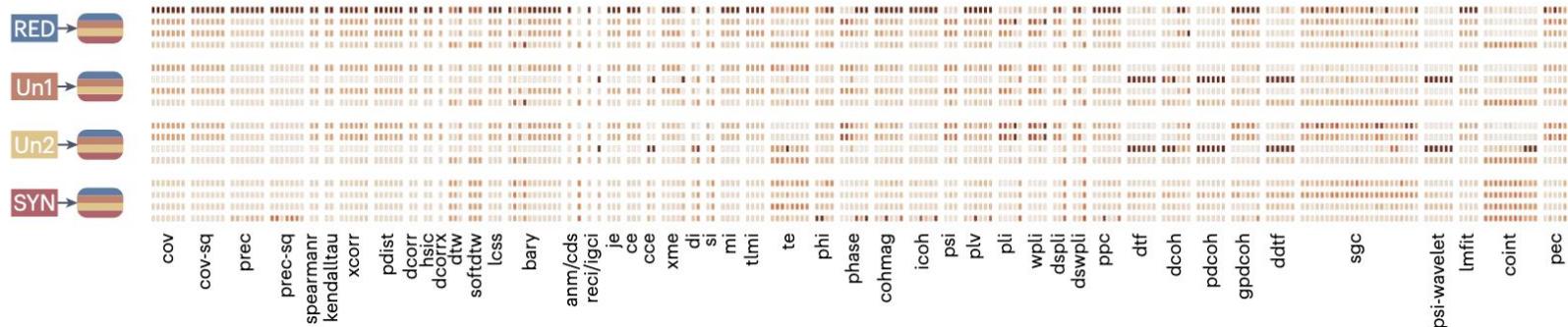


Integrated Information Decomposition

b Information-dynamic atoms

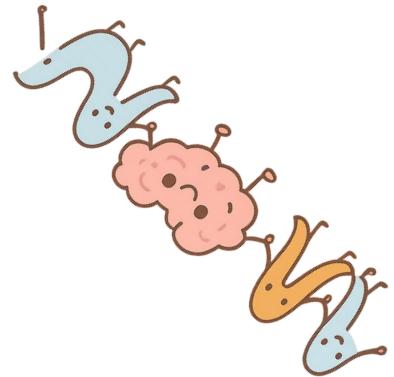
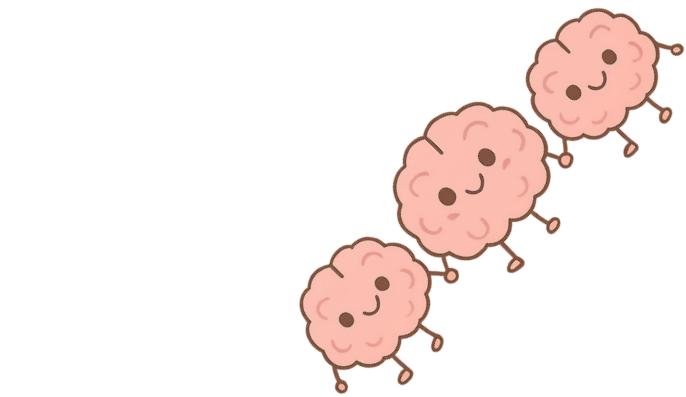
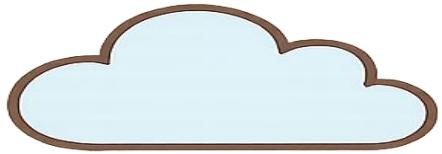


c Relative contribution of information atoms



Discussion

- FCs derived with different statistics look very different and have different hubs.
- They all show expected *weight-distance* and *structure-function* correlations, but with variable magnitude.
- Variability rises from pairwise statistics being sensitive to different underlying mechanisms of interregional signaling.
 - Pairwise statistic should be matched to the experimental question.
 - For fMRI, covariance based methods seem to be most balanced.
 - Precision (i.e., inverse covariance) is surprisingly good for fingerprinting.



The end

