
THE ATTRACTOR STATES OF THE FUNCTIONAL BRAIN CONNECTOME

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Abstract

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Keywords

Highlights:

- continuous "activity flow" across brain regions
- considering the functional brain connectome as the weights of

1 Introduction

Measuring localized brain activity, e.g. with functional magnetic resonance imaging (fMRI) have provided a wealth of insights into the neural correlates of several behavioral/psychological phenotypes and clinical conditions.

However, functional connectivity has been shown to be dynamic; it fluctuates spontaneously over time in a globally coordinated manner. Dynamic functional connectivity has quasi periodic properties (ref), with a limited number of recurring states ("brain states") ([Gutierrez-Barragan et al. \[2019\]](#)); i.e. sporadic intervals during which information can be efficiently exchanged between a characteristic subset of brain regions ([Hutchison et al. \[2013\]](#), [Liu and Duyn \[2013\]](#), [Zalesky et al. \[2014\]](#)). Brain state dynamics can be assessed with multiple techniques, including independent component analysis (ref), co-activation patterns ([Liu and Duyn \[2013\]](#)), and there is accumulating evidence for the neurobiological relevance of these dynamics, with promising perspectives for facilitating the clinical translation of functional neuroimaging techniques by improving brain based biomarkers ([Lee et al. \[2021\]](#)).

However, progress is limited by gaps in the mechanistic understanding of how brain activity and functional connectivity dynamically shape each other and how this process leads to the emergence and organization of brain states (ref).

Here we propose a novel model-based framework that - with minimal and reasonable assumptions about the "activity flow" (ref: Cole-papers) between two, functionally connected regions - considers the stationary functional brain network as an already-trained artificial neural network. In the proposed framework, the topology of the stationary brain connectome defines a cost (energy) for any arbitrary brain activation patterns and a trajectory towards one of the finite number of stable patterns that minimize this cost (so-called attractor states).

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- noise

Here we propose these attractors states as robust and neurobiologically relevant characteristics of the functional brain connectome, with a wide variety of potential applications.

We demonstrate that the proposed attractor states highly resemble to the dynamic brain states commonly observed by dynamic functional connectivity methods (e.g. CAP-analyses (ref)) and provide a proof-of-concept for the biomedical validity of our framework, by showing that the average brain activations corresponding to the attractor states during resting sate display manifold significant associations to cognition.

Due to the known noise-tolerance of the applied eANN-s, the proposed approach can be expected to be highly robust/reliable/replicable, which we demonstrate with independent datasets (total n=xxx).

List all the aims: hierarchy, generalizability etc

2 Results

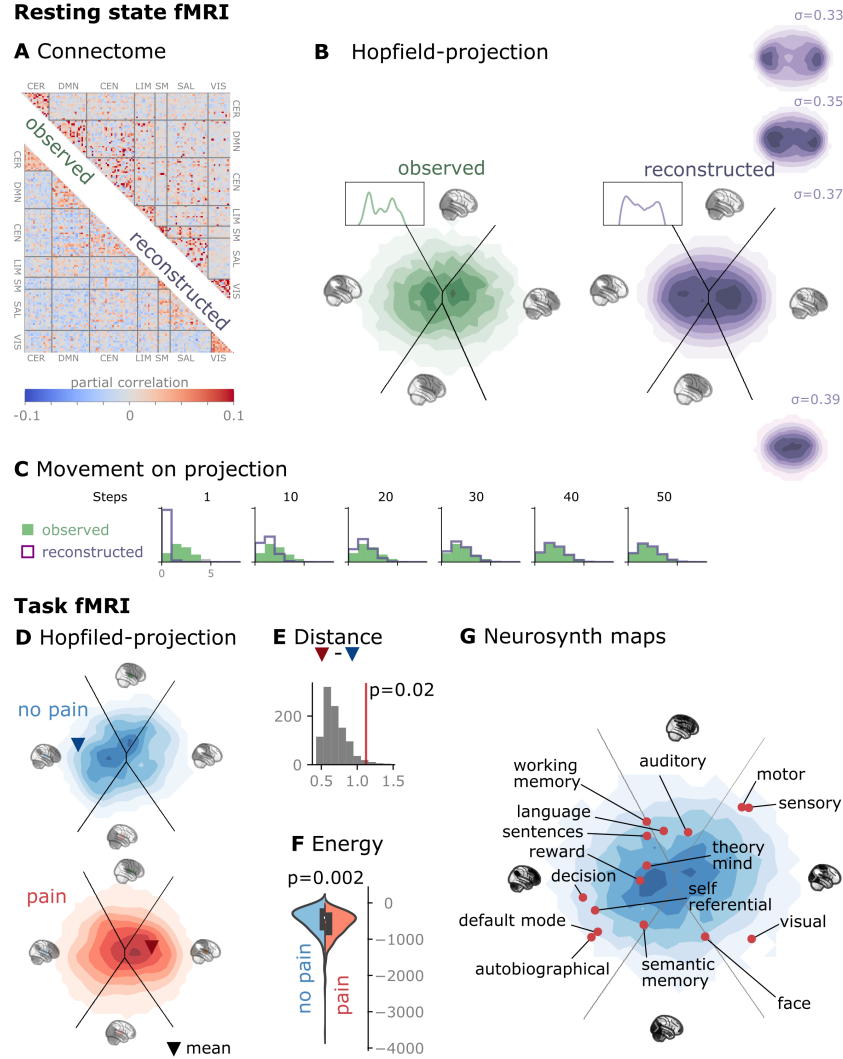


Figure 1: Empirical Hopfield-networks reconstruct real brain activity.

Here I refer to Figure 1.

3 Discussion

4 Methods

Todo

Todo

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