Tournaplexes and their Applications to Neuroscience

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ORIGINAL RESEARCH ARTICLE

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#### Cliques of Neurons Bound into Cavities Provide a Missing Link between Structure and Function

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The lack of a formal link between neural network structure and its emergent function has hampered our understanding of how the brain processes information. We have now come closer to describing such a link by taking the direction of synaptic transmission into account, constructing graphs of a network that reflect the direction of information flow, and analyzing these directed graphs using algebraic topology. Applying this approach to a local network of neurons in the neocortex revealed a remarkably intricate and previously unseen topology of synaptic connectivity. The synaptic network contains an abundance of



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- $\bullet~\sim 8\,000\,000$  Connections
- 6 Layers
- 54 Neuronal Types
- Functional Model



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#### A Directed Graph





#### Study the structure of the brain graph

and

# Stimulate the circuit and use the resulting activity to determine which stimulus was applied

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#### Transitive Cliques $\longrightarrow$ Simplices

(i.e those containing no directed cycles)

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#### Structure



# Function



 $\mathsf{Directed}\ \mathsf{Graph} \longrightarrow \mathsf{Geometric}\ \mathsf{Realisation}\ \mathsf{of}\ \mathsf{Simplicial}\ \mathsf{Set}$ 

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Directionality of clique  $\sigma$ :

$$dr_{\textit{local}}(\sigma) = \sum_{v \in \sigma} (\textit{indegree}_{\sigma}(v) - \textit{outdegree}_{\sigma}(v))^2$$

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$$dr_{\textit{local}}(\sigma) = \sum_{\mathsf{v} \in \sigma} (\textit{indegree}_{\sigma}(\mathsf{v}) - \textit{outdegree}_{\sigma}(\mathsf{v}))^2$$

$$dr_{global}(\sigma) = \sum_{v \in \sigma} (indegree_G(v) - outdegree_G(v))^2$$

































## **Directionality Filtrations**





# Distinguishing Graphs Using Tournaplex

Data: 200 Erdős-Rényi graphs with edge  $i \rightarrow j$  present with probability  $\begin{cases} 0.25, & \text{if } i < j \\ q, & \text{if } i > j \end{cases}$ 



# Distinguishing Stimuli

Data: 45 spike trains on Blue Brain model, 5 repetitions of 9 different stimuli



# Thanks for Listening!

Many Thanks to the Blue Brain Project

Complexes of Tournaments, Directionality Filtrations and Persistent Homology arXiv:2003.00324