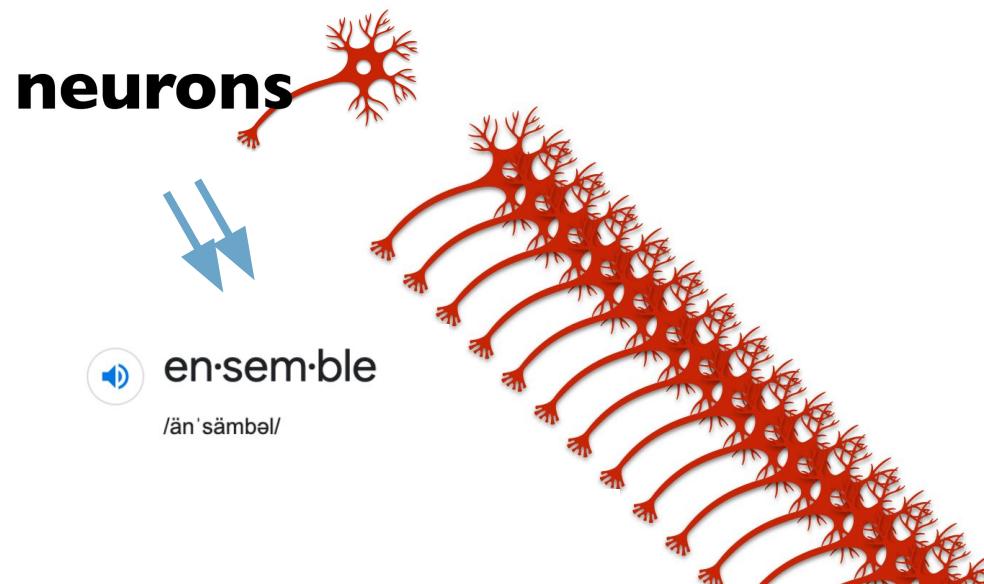


Structure and Function of A Mouse Brain

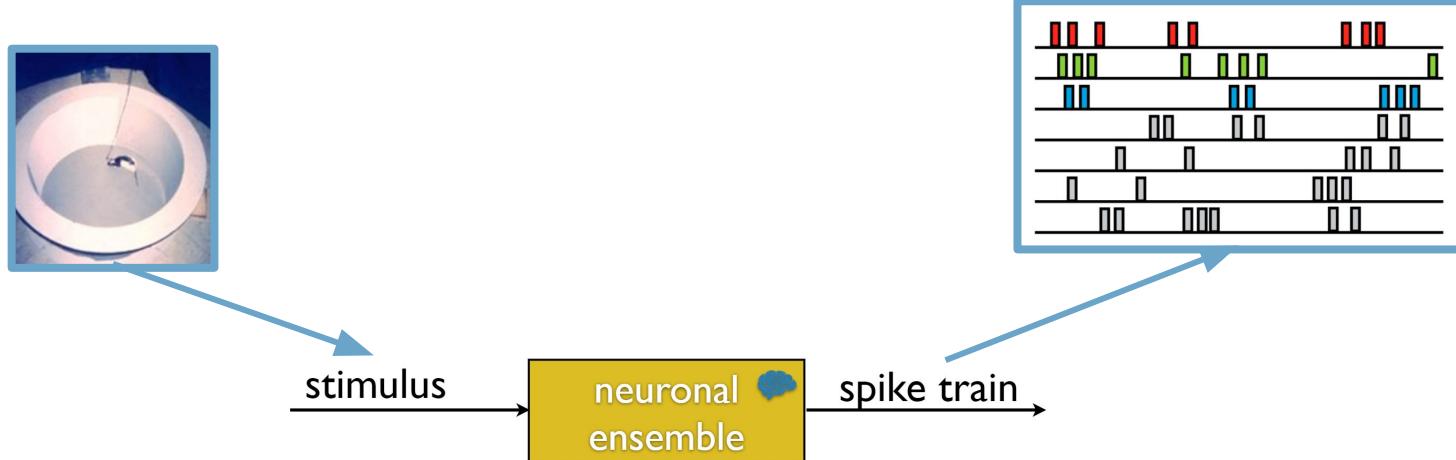
*Cliques of Neurons Bound into Cavities Provide a Missing Link
between Structure and Function* Michael W. Reimann et al.

— Blue Brain Project
12 June 2017 **EPFL**

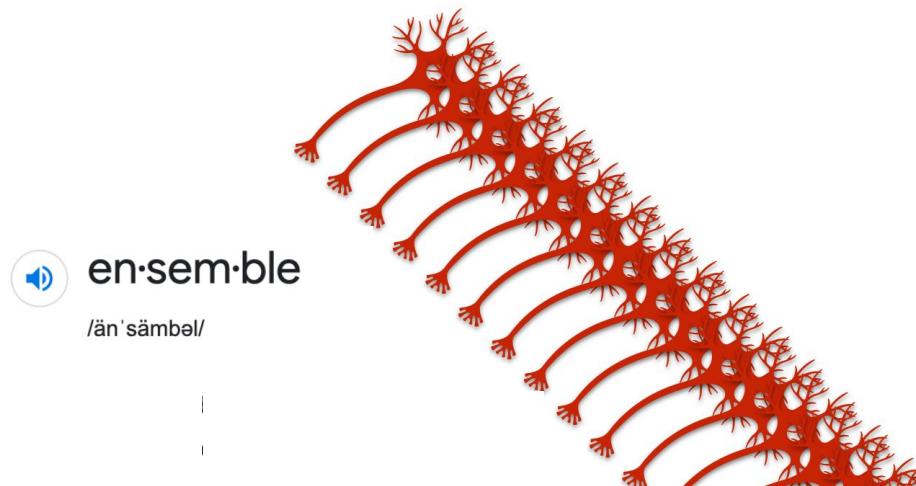
Structure & Function?



Structure & Function?

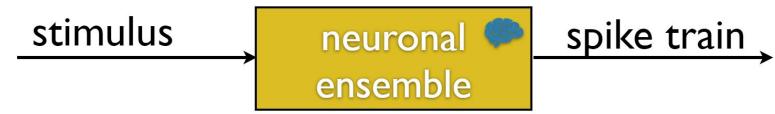


Structure & Function?



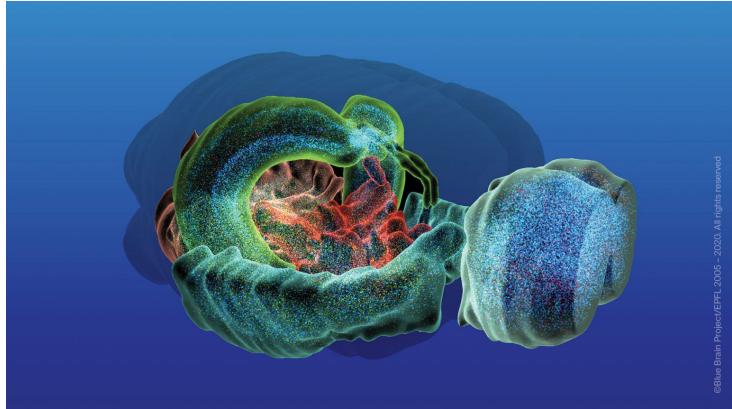
en·sem·ble

/än'sämbəl/



Blue Brain Project

- Henry Markram (EPFL), founded in 2005
- Biologically detailed digital reconstructions and simulations of the mouse brain
 - Cellular model (2005)
 - Neural cliques (2017)
 - Digital 3D brain atlas (2018)



Contents

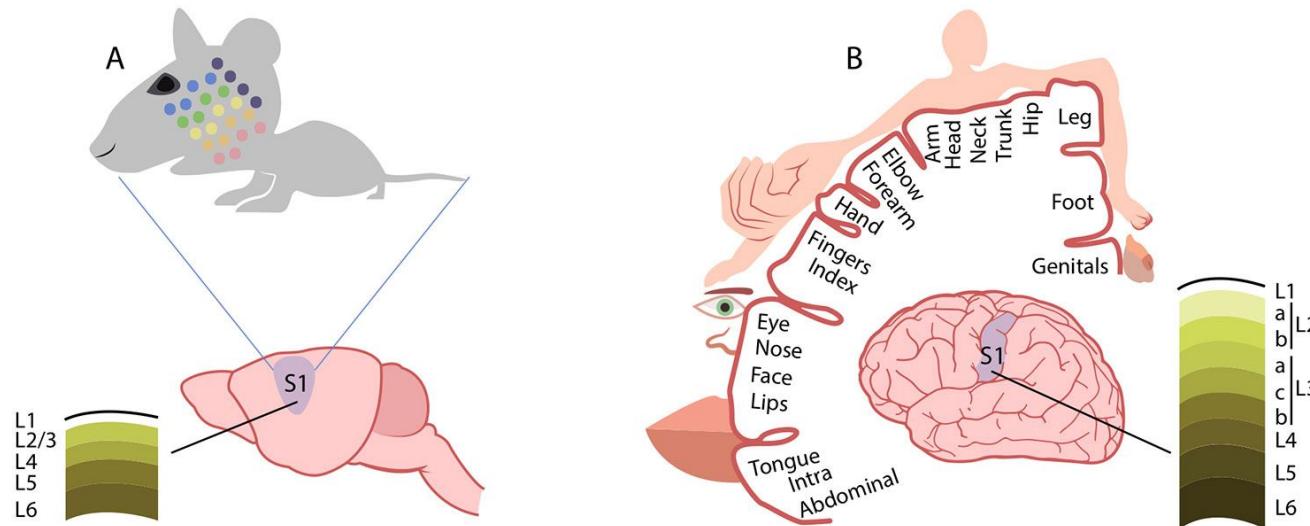
Structure (Cliques of Neurons)

- Local
 - Biological background
 - Topological Construction
 - Directed Graph
 - Simplices & Complexes
- Global

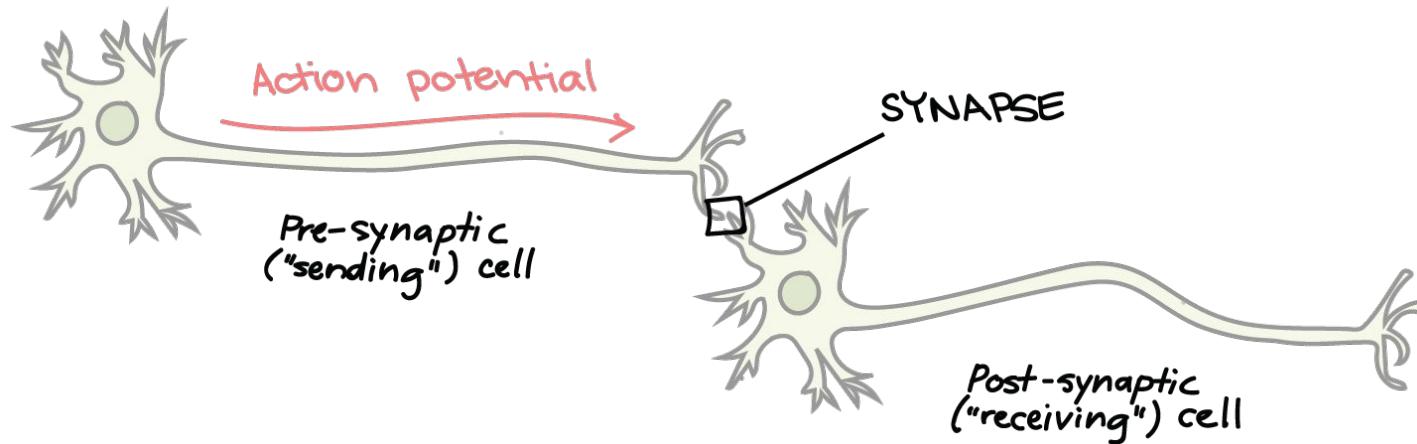
Function (Bound into Cavities)

- Local
- Global

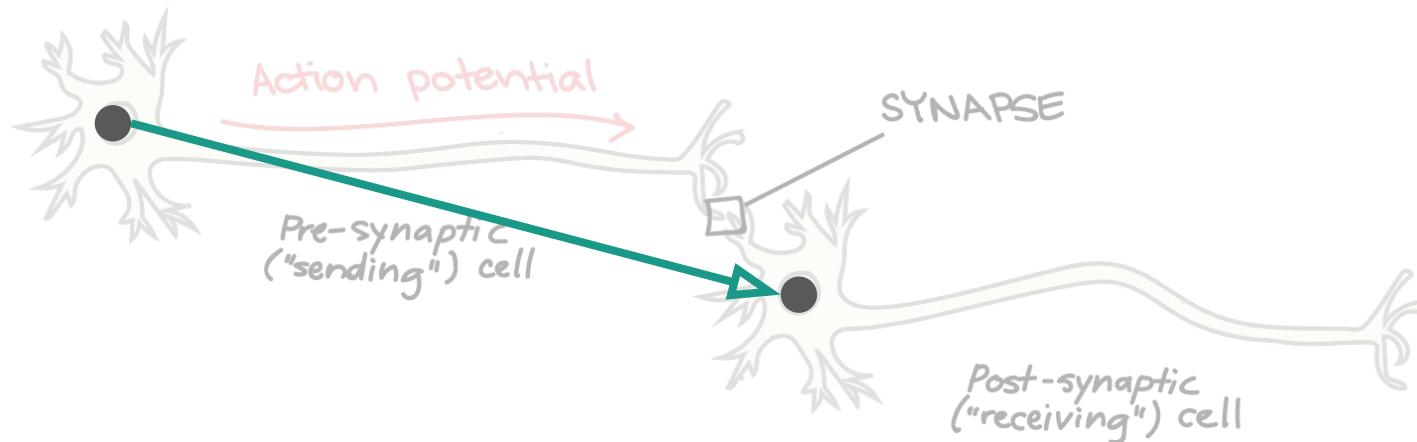
Primary Somatosensory Cortex



Neurons & Synapse



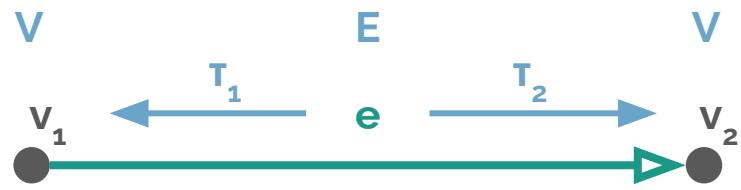
Directed Graphs



Directed Graphs \mathcal{G}

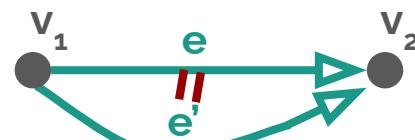
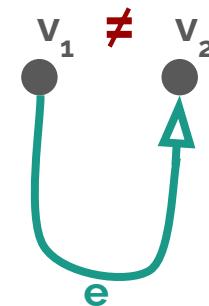
- Finite sets (V, E)
- $\tau = (\tau_1, \tau_2) : E \rightarrow V \times V$
- $\tau_1(e) = v_1$, source vertex
- $\tau_2(e) = v_2$, target vertex

direction



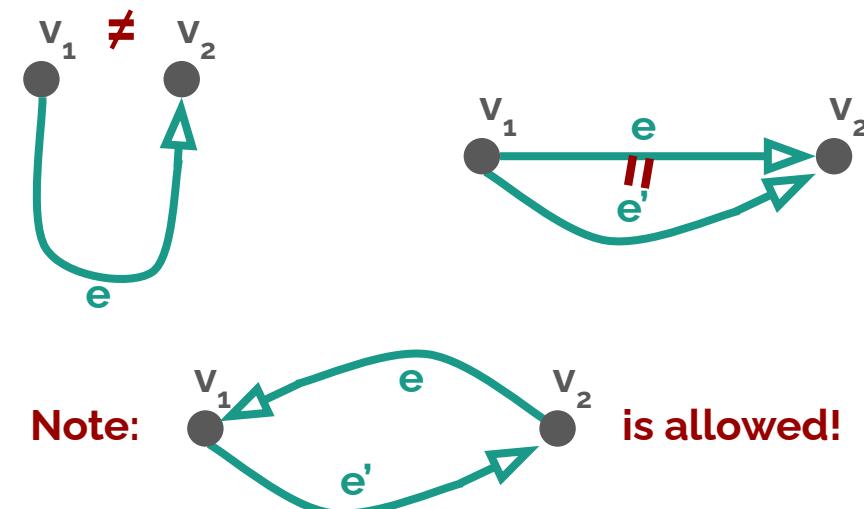
Directed Graphs - function τ

- 1) No self-loop
- 2) Injective



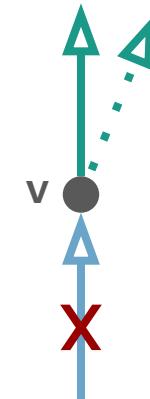
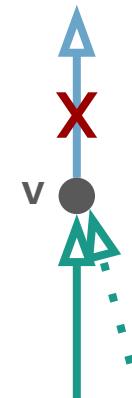
Directed Graphs - function τ

- 1) No self-loop
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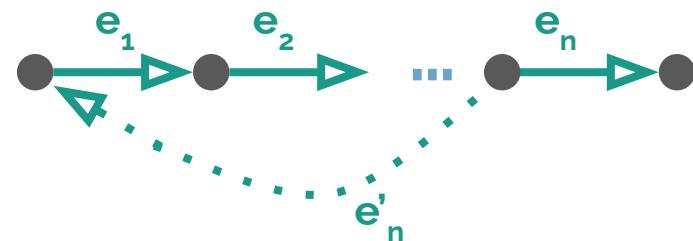
Directed Graphs - vertex v

- Sink
 - if there exists no $e \in E$ such that $v = \tau_1(e)$, but there is at least one edge $e' \in E$ such that $\tau_2(e')=v$
- Source
 - if there exists no $e \in E$ such that $v = \tau_2(e)$, but there is at least one edge $e' \in E$ such that $\tau_1(e')=v$



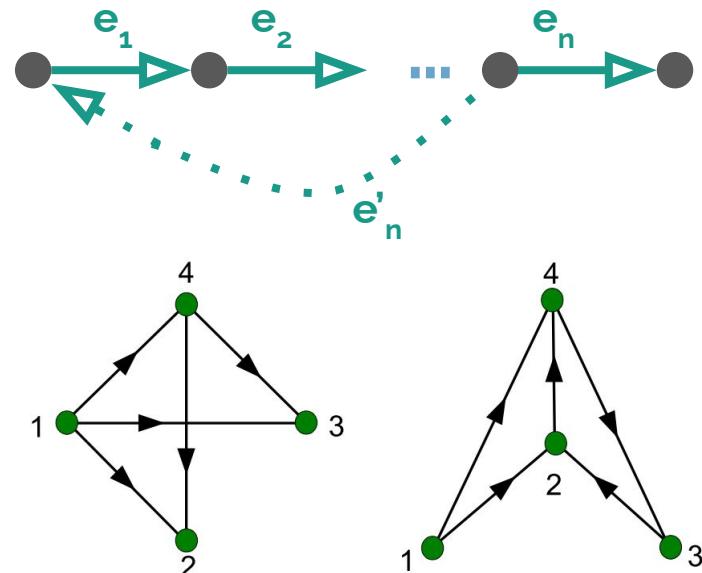
Directed Graphs - path

- Sequence of edges (e_1, \dots, e_n)
 - $\tau_2(e_k) = \tau_1(e_{k+1})$ for all $1 \leq k < n$
- Length := n
- Oriented circle
 - $\tau_2(e_n) = \tau_1(e_1)$



Directed Graphs - path

- Sequence of edges (e_1, \dots, e_n)
 - $\tau_2(e_k) = \tau_1(e_{k+1})$ for all $1 \leq k < n$
- Length := n
- Oriented circle
 - $\tau_2(e_n) = \tau_1(e_1)$
- Graph containing no oriented circle is acyclic.



Abstract directed simplex & simplicial complex

- An abstract direct simplicial complex is a collection \mathcal{S} of finite, ordered sets which is closed under taking subsets.
- A subcomplex is a subcollection of \mathcal{S} that is itself an abstract directed simplicial complex.
- Elements of \mathcal{S} are called its simplices, denoted σ .

Abstract directed simplex & simplicial complex

- Dimension of σ , denoted $\dim(\sigma) := (\text{cardinality of } \sigma) - 1$.
- σ is called an n-simplex if it has dimension n.
- Set of all n-simplices is denoted \mathcal{S}_n .

Abstract directed simplex & simplicial complex

- τ is a face of $\sigma = (v_0, \dots, v_n)$ if it is a subset with strictly smaller cardinality.
- front face = (v_0, \dots, v_m) , $0 < m < n$
- back face = (v_m, \dots, v_n) , $0 < m < n$
- i-th face of σ is the $(n-1)$ -simplex obtained from σ by removing vertex v_{n-i}
- A simplex is maximal if it is not a face of any other simplex.

Abstract directed simplex & simplicial complex

Remark 1

The set of all maximal simplices of a simplicial complex determines the entire simplicial complex, since every simplex is either maximal itself or a face of a maximal simplex.

(realized) **Simplicial complex**

- Simplicial complex is induced from an abstract simplicial complex by geometric realization. It gives rise to a topological space.
 - 0-simplex: a point
 - 1-simplex: an edge
 - 2-simplex: a filled-in triangle
 - ...
- Simplicial complexes are realized by gluing realized simplices together by their common faces.

(realized) **Simplicial complex**

Remark 2

If for two simplices, neither of them is a face of the other, then their intersection is a proper subset, therefore a face, of both simplices.

Thus, simplicial complexes are well-defined geometric objects.

Directed flag complex of \mathcal{G}

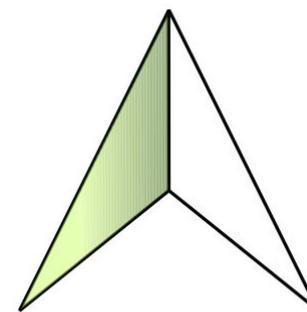
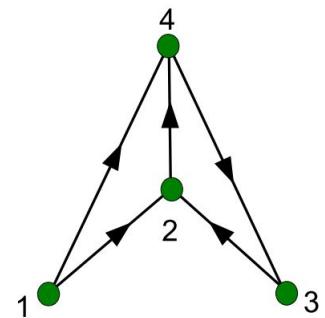
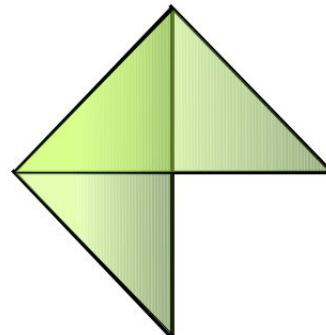
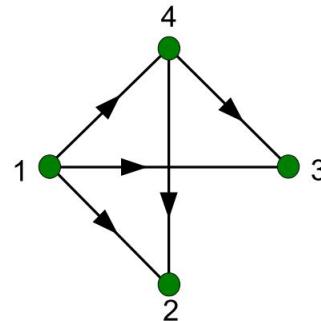
- For $\mathcal{G} = (V, E, \tau)$ a directed graph, the associated directed flag complexes are defined by $\mathcal{S} = \mathcal{S}(\mathcal{G})$ that
 - \mathcal{S}_0 (set of all 0-simplices) = V
 - $\mathcal{S}_n = \{(v_0, \dots, v_n)\}$ such that there exists an edge directed from v_i to v_j for every $0 \leq i < j \leq n$
 - Therefore v_0 is the source and v_n is the sink of the simplex.

Directed flag complex of \mathcal{G}

Remark 3 (disambiguation)

An n -simplex in \mathcal{S} is characterized by the (ordered) sequence (v_0, \dots, v_n) , but not by the labeling of vertices. For instance (v_1, v_2, v_3) and (v_2, v_1, v_3) are **distinct** 2-simplices with the same set of vertices.

Directed flag complex of \mathcal{G}



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- Global
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 - Construction
 - Robustness check

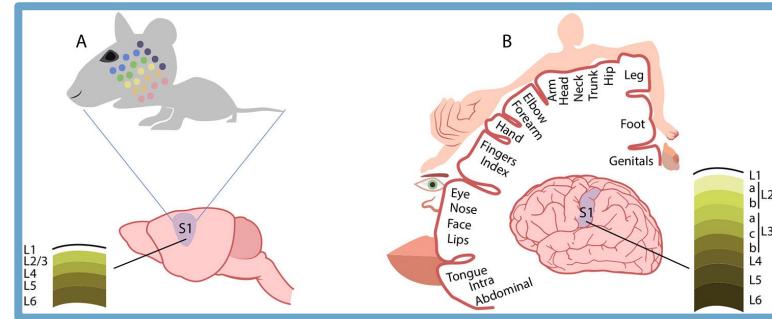
Function (Bound into Cavities)

- Local
- Global

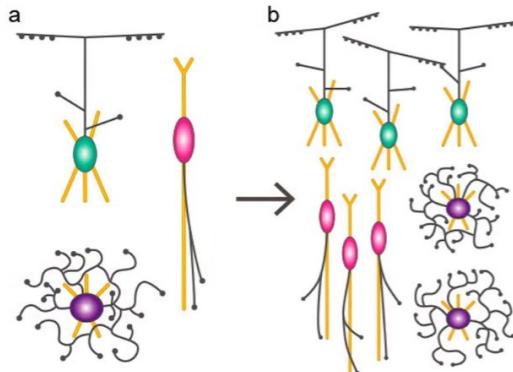
Structure - Biological Validation

'Core' Sample

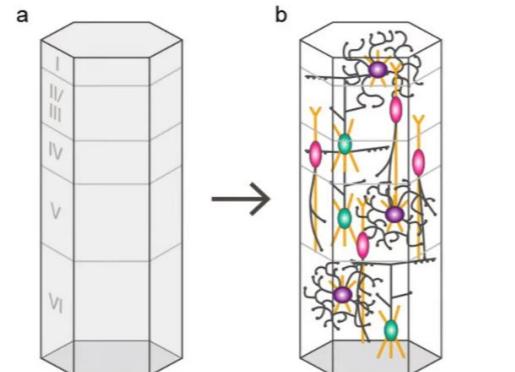
- 5 samples: 2mm high, 0.25mm radius



A Morphological diversity of neurons:
(a) m-types, (b) cloning



B Microcircuit anatomy: (a) Microcircuit dimensions,
(b) m-type distribution, and morphology selection



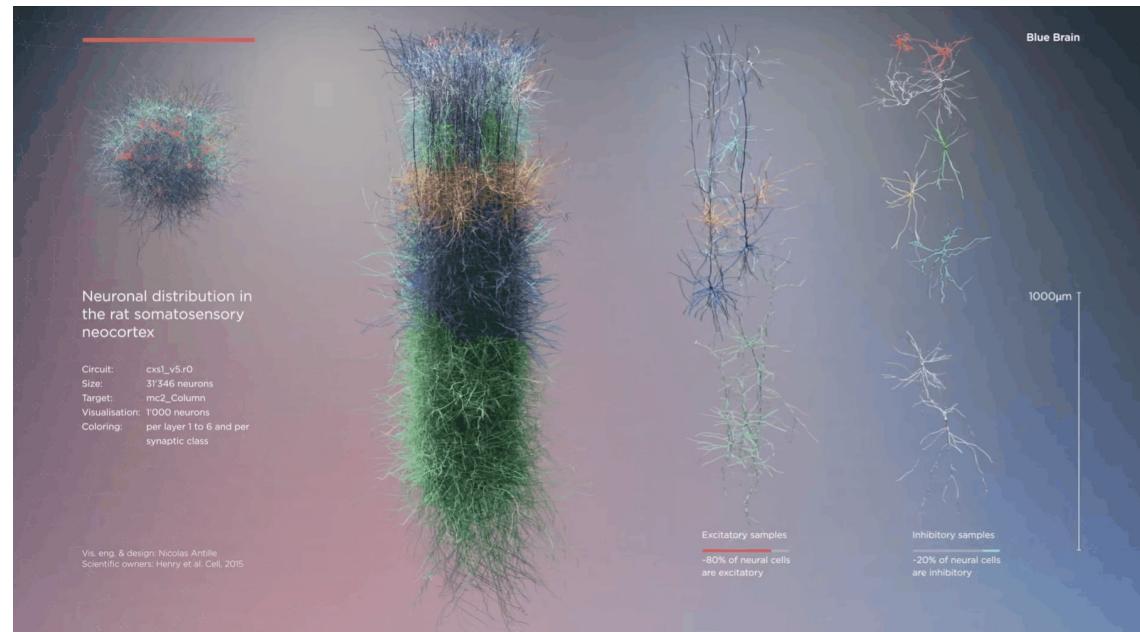
C Reconstructing
microcircuit connectivity



Structure - Digital Reconstruction

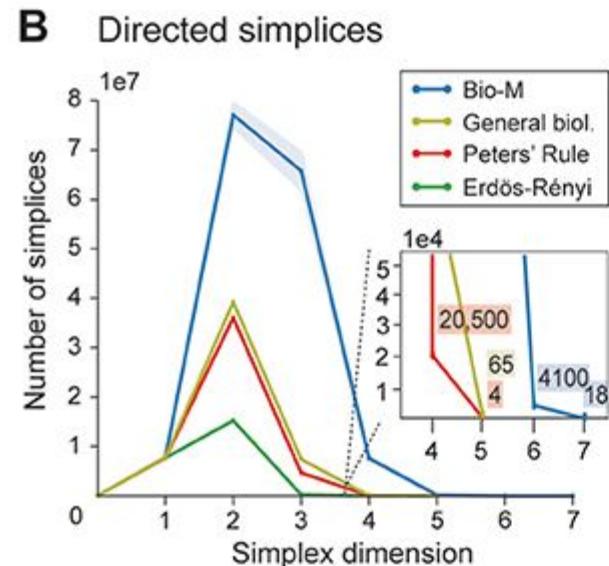
Microcircuit

- ~8 million connections (edges)
- ~31,000 neurons (vertices)



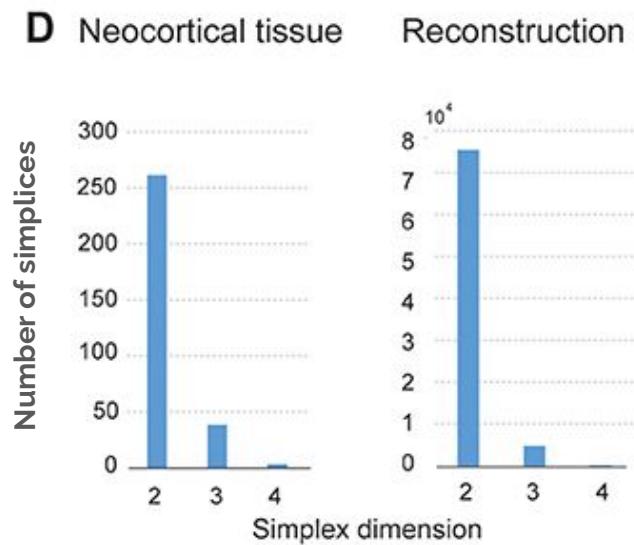
Abundance of Directed Simplices

- Mean of the 5 microcircuits based on samples from 5 mice
- Compared to circuits with the same size that
 - Preserve distance dependance
 - Preserve 3D neuronal model
 - Is connected with the same average connection probability (~0.8%)



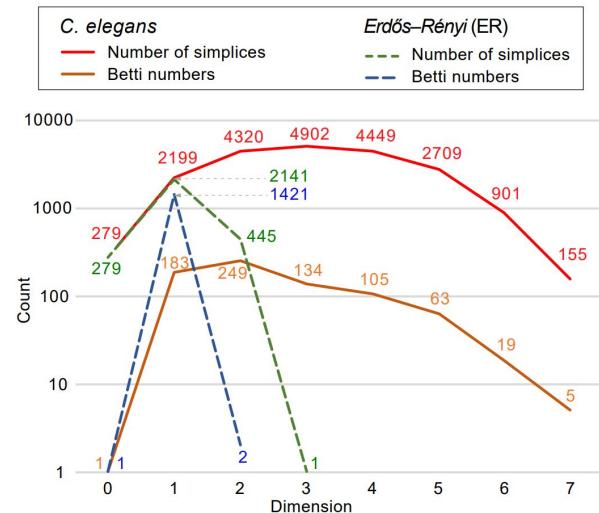
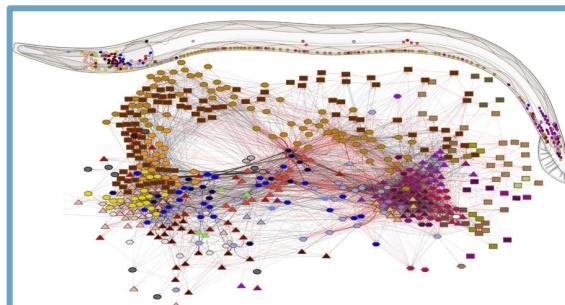
Robustness check - Observation

- 55 *in vitro* patch-clamp experiments
 - 6 - 12 neurons, 300 μm slices
 - same age and brain region as samples used for digital reconstruction
- *in silico* simulation
 - volume of $200 \times 200 \times 20 \mu\text{m}$
 - “*repeating the same multi-neuron patch-clamp recordings*”



Robustness check - Neural network of another species

- *C. elegans* connectome
 - a free-living transparent nematode about 1 mm in length that lives in temperate soil environments



Results



Result 1

- The numbers of high-dimensional cliques and cavities found in the reconstruction are far **higher than** in null models, even in those closely resembling the biology-based reconstructed microcircuit
- Their existence is **verified** in actual neocortical tissue
- **Similar structures** are found in a nervous system as phylogenetically different as that of the worm *C. elegans*

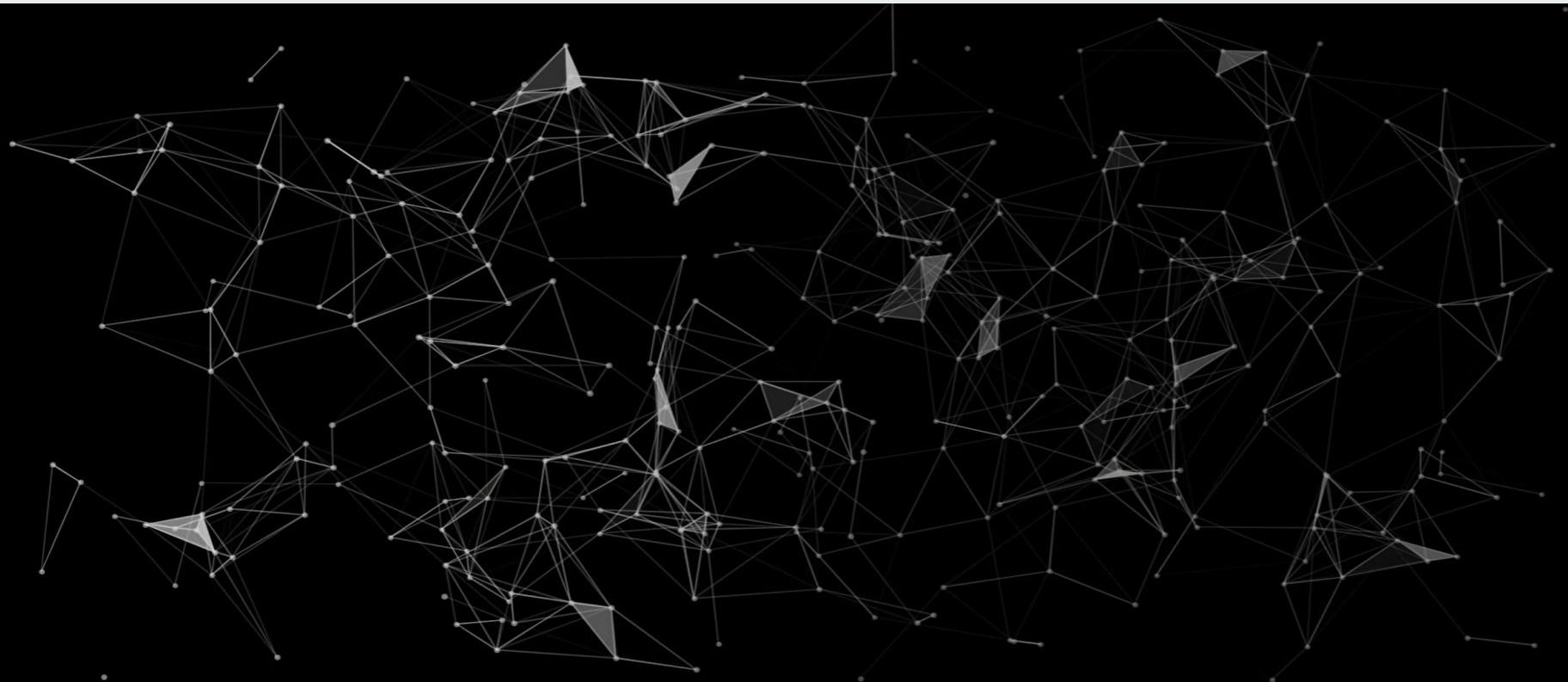
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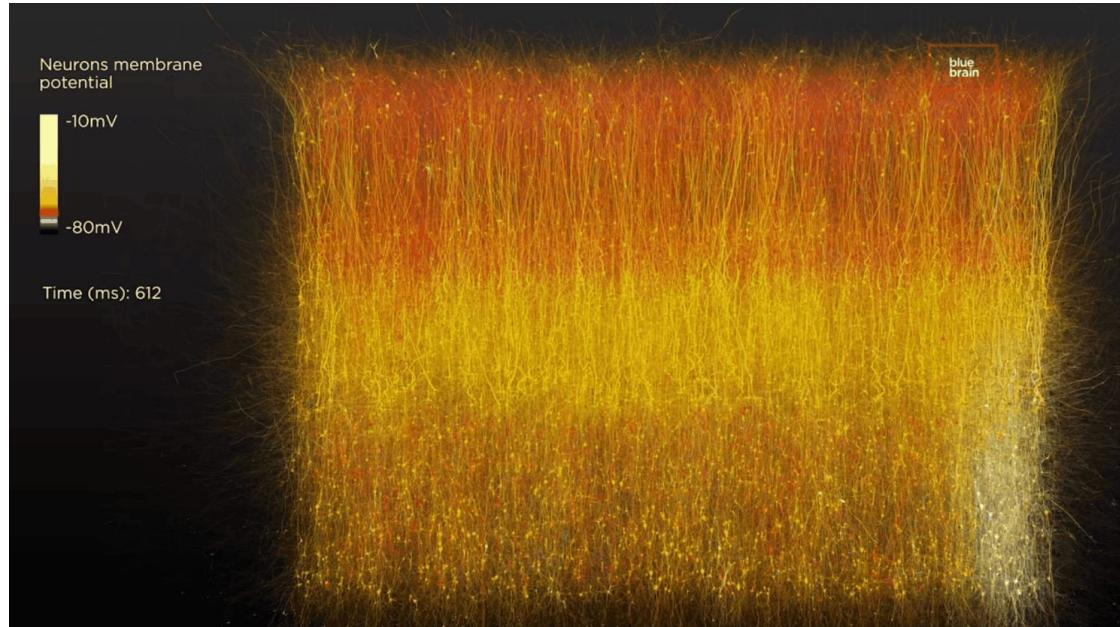
Function (Bound into Cavities)

- Local
- Global



Break...

Synaptic reaction to stimuli (*in vivo*)



Contents

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 - *Complexes: Homology?*

Function (Bound into Cavities)

- Local
 - Simplices
 - Topological features
 - Spiking response
- Global

Directionality of finite directed graphs

- For each vertex $v \in \mathcal{G}$, define the signed degree of v to be

$$sd(v) = \text{Indeg}(v) - \text{Outdeg}(v). \quad (1)$$

- Then, define the directionality of \mathcal{G}

$$\text{Dr}(\mathcal{G}) = \sum_{v \in V} sd(v)^2. \quad (2)$$

Directionality of finite graphs

Remark 4

If \mathcal{G}_n is a directed n-simplex, let \mathcal{G} be any directed graph on $(n+1)$ vertices, then

$$\text{Dr}(\mathcal{G}) \leq \text{Dr}(\mathcal{G}_n).$$

If additionally \mathcal{G} is a fully connected directed graph without reciprocal connections (n -simplex), then the equality holds if and only if the two directed graphs are isomorphic to each other.

Recall...

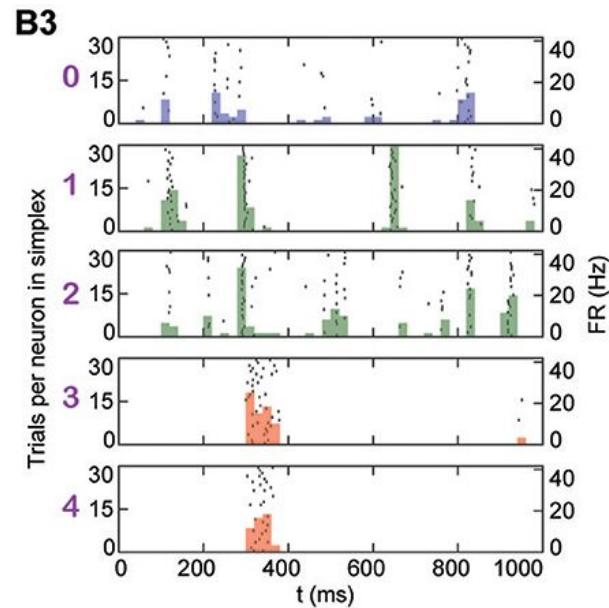
Abstract directed simplex & simplicial complex

Remark 1

The set of all maximal simplices of a simplicial complex determines the entire simplicial complex, since every simplex is either maximal itself or a face of a maximal simplex.

Average Spiking Response

- peri-stimulus time histogram (PSTH)
 - bin size: 25 ms
 - 5 seconds of thalamo-cortical input over 30 trials of a given input pattern



Spike Correlations

- Concatenating PSTH of 9 different thalamic stimuli of 5 seconds, 30 trials
- The normalized covariance matrix of the PSTHs of all neurons is calculated by

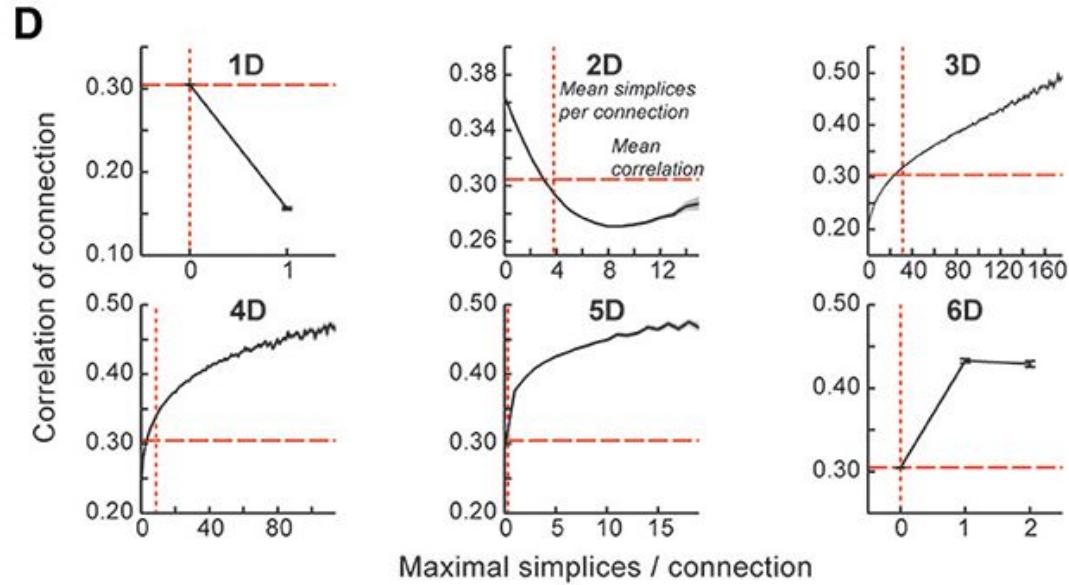
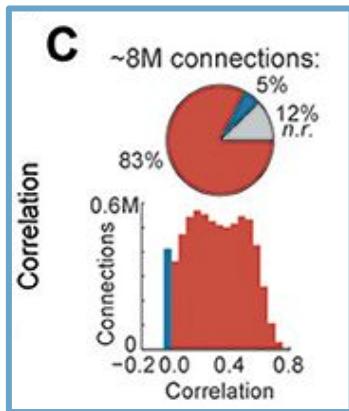
$$R_{ij} = \frac{C_{ij}}{\sqrt{C_{ii}C_{jj}}}, \quad (9)$$

where C_{ij} is the covariance of the PSTHs of neurons i and j.

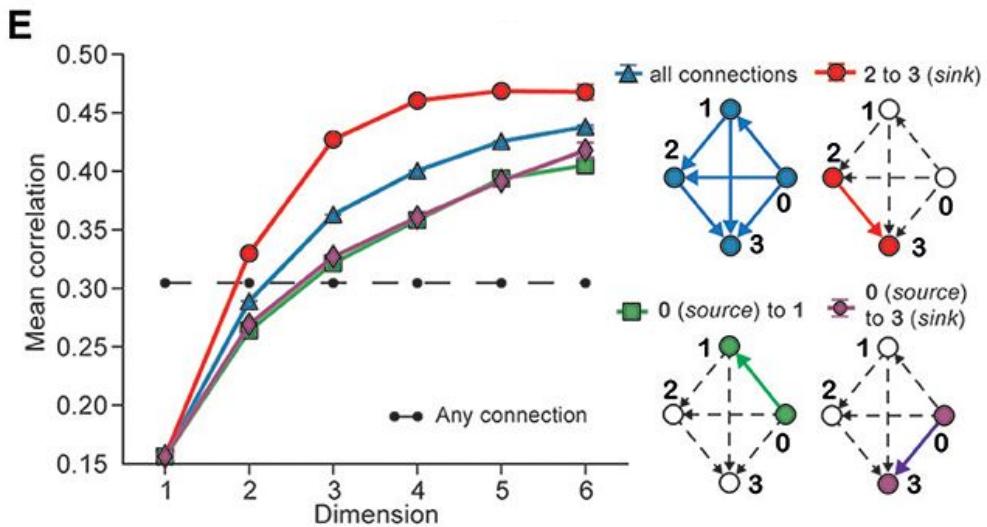
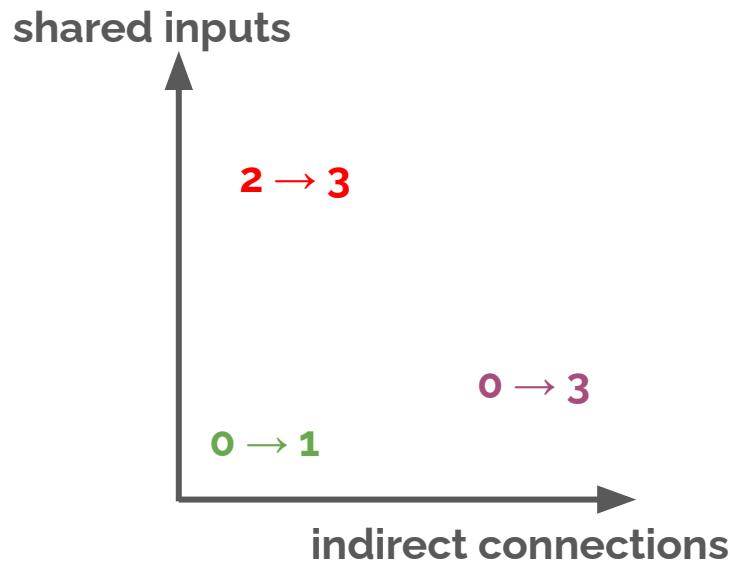
Remark 5

Each entry of R_{ij} refers to a particular connection in at least one maximal simplex.

Spike Correlations



Spike Correlations - pairwise



Results



Result 2

- **Spike correlation of a pair of neurons** strongly increases with the **number and dimension** of the **cliques** they belong to
- It even depends on their **specific position** in a directed clique
- Generalize the similar link for motifs built from 2-dimensional simplices found by previous studies to higher dimensions

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Function (Bound into Cavities)

- Local
 - Simplices
 - Topological features
 - Spiking response
- Global
 - Complexes: Homology
 - Betti number & Euler characteristic
 - Filtration
 - Construction
 - Brief evaluation

Betti Number & Euler characteristic

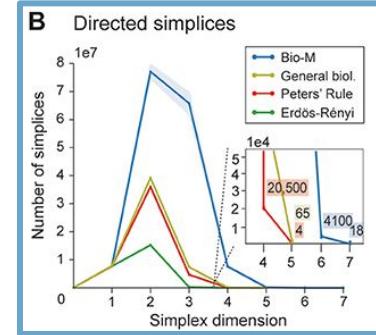
- Homology groups are computed using mod 2 coefficients, notation: \mathbb{F}_2 or $\mathbb{Z}/2\mathbb{Z}$
- The n-th Betti Number indicates the number of “n-dimensional cavities” in \mathcal{S}

$$\beta_n(\mathcal{S}) = \dim(H_n(\mathcal{S}))$$

- The Euler characteristic is defined to be

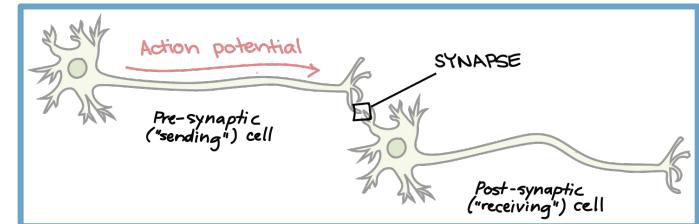
$$\chi(\mathcal{S}) = \sum_{n \geq 0} (-1)^n \beta_n(\mathcal{S}).$$

However...



- The complexity of computing the n -th Betti numbers scales with the number of simplices in dimensions $n - 1$, n , and $n + 1$. (*Rank and nullity of matrices of dimension $(n - 1) \times n$ and $n \times (n + 1)$*)
- Calculation of Betti numbers of dimension **above 0 and below 5** was computationally not viable.
- Euler characteristic computations imply that at least one of β_2 or β_4 must be nonzero, and it is highly likely the β_k is nonzero for all $k \leq 5$.

Transmission Response Matrices

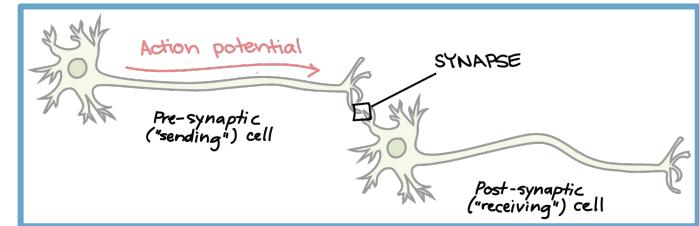


- Want to construct time series of sub-graphs with all vertices and *active edges*.
- Define the temporal sequence of transmission-response matrices associated to a simulation of neuronal activity of duration T

$$TR(\Delta t_1, \Delta t_2) := \{A(n) = A(n, \Delta t_1, \Delta t_2)\}_{n=1}^N, \quad (10)$$

where n-th matrix $A(n)$ is a binary matrix recording spiking activities within time interval $[n \cdot \Delta t_1, (n + 1) \cdot \Delta t_1 + \Delta t_2]$, and where $N = T/\Delta t_1$.

Transmission Response Matrices

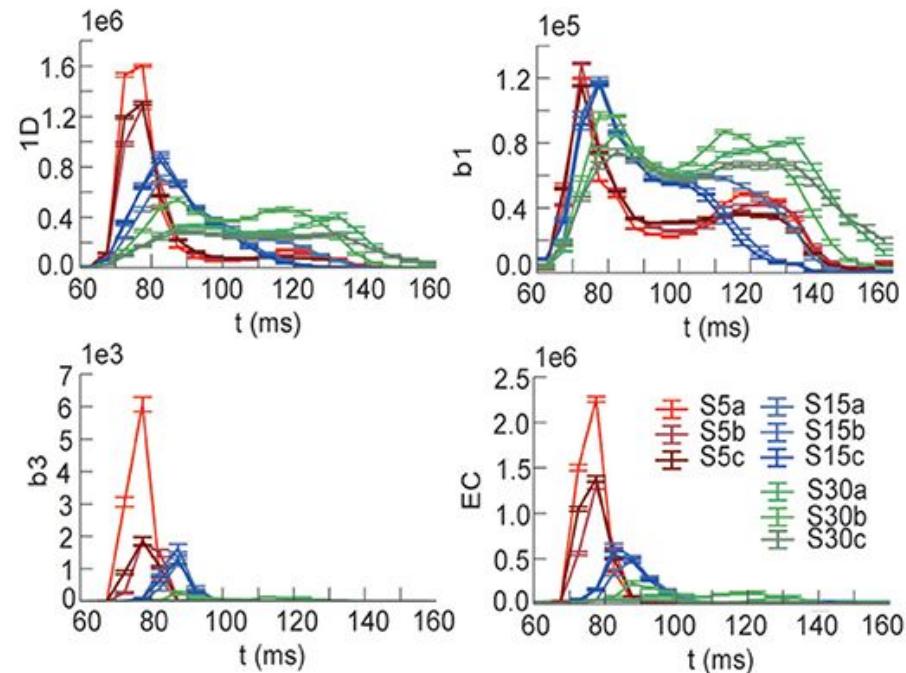


$$TR(\Delta t_1, \Delta t_2) := \{A(n) = A(n, \Delta t_1, \Delta t_2)\}_{n=1}^N, \quad (10)$$

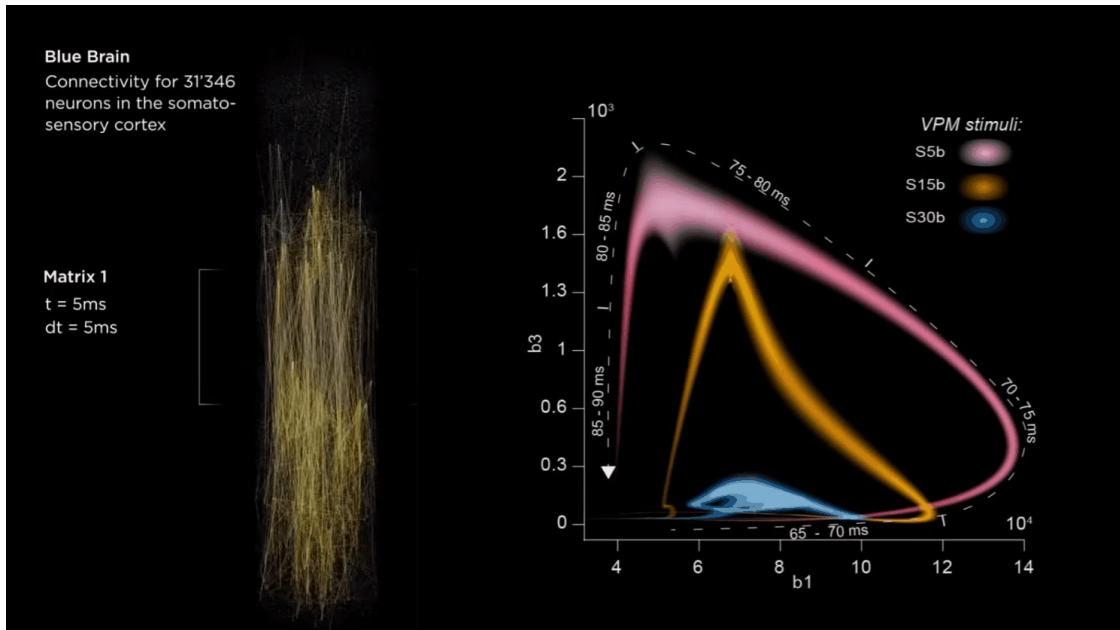
- $A(n)_{ij}$ is 1 if and only if
 - neuron j to neuron k forms a pre-post synaptic pair. (*Namely, there exists an edge between the two corresponding vertices regardless of direction.*)
 - neuron j spikes in the n-th time bin
 - neuron k spikes after neuron j, within a Δt_2 interval
- $\Delta t_1 = 5$ and $\Delta t_2 = 10$ are optimized parameter to best reflect the actual successful transmission of signals between the neurons in the microcircuit.

Bound into Cavities

- neurons become bound into cliques and cavities by correlated activity

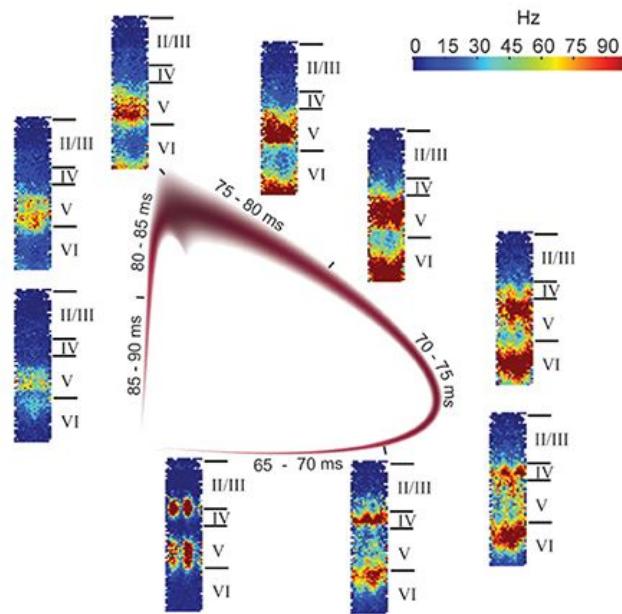


Bound into Cavities

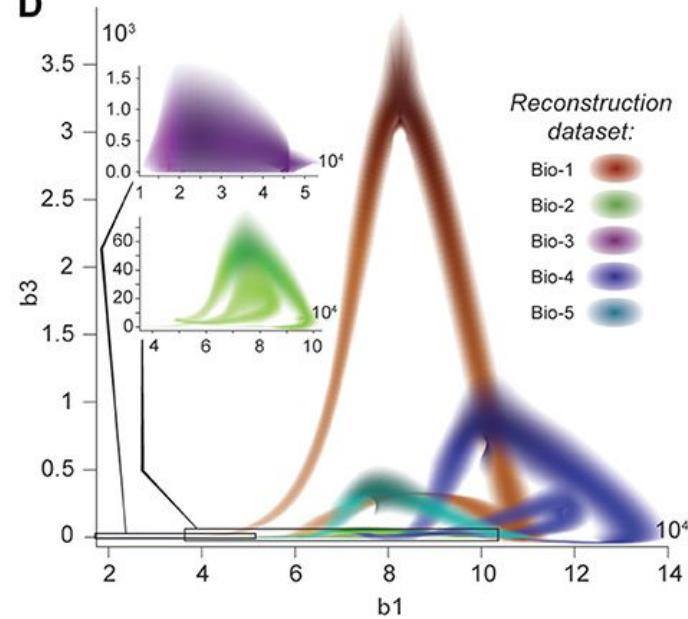


Bound into Cavities

C



D



Results

Result 3

- Applying **topological metrics** (homology) to time-series of transmission-response sub-graphs revealed a **sequence of cavity formation and disintegration** in response to stimuli, consistent across different stimuli and individual microcircuits
- The size of the trajectory was determined by the degree of synchronous input and the biological parameters of the microcircuit, while its location depended mainly on the biological parameters
- Neuronal activity is therefore organized not only within and by directed cliques, but also by highly structured relationships between directed cliques, consistent with a recent hypothesis concerning the relationship between structure and function

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 - Topological features
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- Global
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 - Betti number & Euler characteristic
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 - Construction
 - Brief evaluation

Conclusion

This study...

- provides a simple, powerful, parameter-free, and unambiguous mathematical framework for **relating the activity of a neural network to its underlying structure**, both locally (in terms of simplices) and globally (in terms of cavities formed by these simplices)
- suggests that neocortical microcircuits process information through a stereotypical progression of **clique and cavity formation and disintegration**, consistent with a recent hypothesis of common strategies for information processing across the neocortex

References

- *Cliques of Neurons Bound into Cavities Provide a Missing Link between Structure and Function.* <https://www.frontiersin.org/articles/10.3389/fncom.2017.00048/full>
- *Digital Neuroscientist of the Future | Kathryn Hess Bellwald | TEDxLugano.* <https://www.youtube.com/watch?v=uQKrCKy3h1E>
- *Sensory Abnormalities in Autism Spectrum Disorders: A Focus on the Tactile Domain, From Genetic Mouse Models to the Clinic.* <https://www.pnas.org/doi/10.1073/pnas.1803274115>
- *Redundancy in synaptic connections enables neurons to learn optimally.* <https://www.pnas.org/doi/10.1073/pnas.1803274115>

Supplementary Materials

- *Scientists discover hidden patterns of brain activity.*
<https://actu.epfl.ch/news/scientists-discover-hidden-patterns-of-brain-activ/>
- *Blue Brain solves a century-old neuroscience problem.*
<https://actu.epfl.ch/news/blue-brain-solves-a-century-old-neuroscience-probl/>
- *The Mind-Boggling Math That (Maybe) Mapped the Brain in 11 Dimensions.*
<https://www.wired.com/story/the-mind-boggling-math-that-maybe-mapped-the-brain-in-11-dimensions/>
- *The human brain sees the world as an 11-dimensional multiverse.*
<https://nypost.com/2017/06/13/the-human-brain-sees-the-world-as-an-11-dimensional-multiverse/>
- *Digital 3D atlas of a mouse brain.* <https://bbp.epfl.ch/nexus/cell-atlas/>