

Measuring the functional complexity of nanoscale connectomes: polarity matters

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Summary

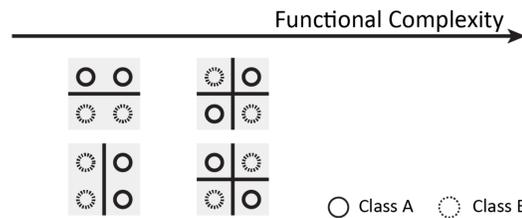
- Why do we have so many excitatory neurons? We propose a functional complexity measurement that is task-agnostic and learning-independent to answer.
- Leveraging EM connectomes, we show that over-abundance of Exc. neurons lead to higher functional complexity, but only when Inhib. neurons are highly connected.
- The predicted E-I properties of highest functional complexity match the real brains, providing a normative explanation to these highly conserved E-I connectivity properties.
- These insights learnt from whole-brain nanoscale connectomes are further leveraged to provide clues to guide the development of deep neural networks.

Motivation

How does a network's Excitatory-Inhibitory connectivity structure affect its functional complexity?

Functional Complexity

Functional complexity = fraction of 'XORable' subnetworks



The two XOR type tasks on the right are *not* linearly-separable, thus more complex than the two on the left.

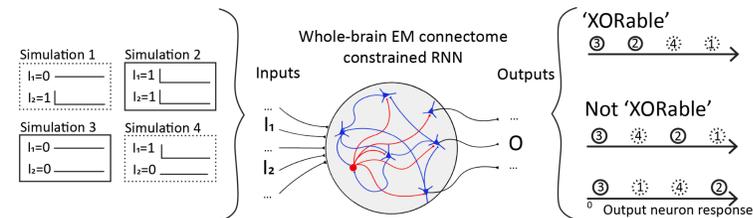
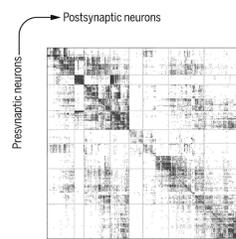


Illustration on the procedure of experimentally determining whether a subnetwork, defined by $\{I_1, I_2, O\}$, is XORable or not.

EM connectome constrained model



(Winding et al., 2023)
(Dorkenwald et al., 2024)

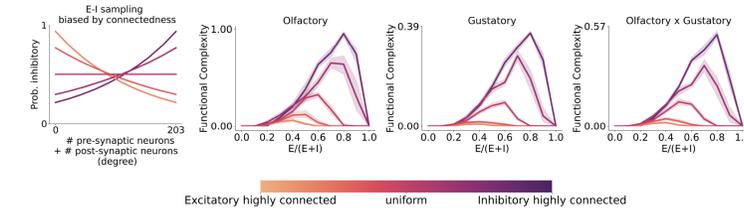
Whole-brain RNNs

$$\mathbf{r}(t) = \text{ReLU}(\mathbf{W}^T \mathbf{r}(t-1) + \mathbf{E}(t))$$

neurons $n_{\text{Larva Drosophila}} = 2952,$
 $n_{\text{Adult Drosophila}} = 132,490$

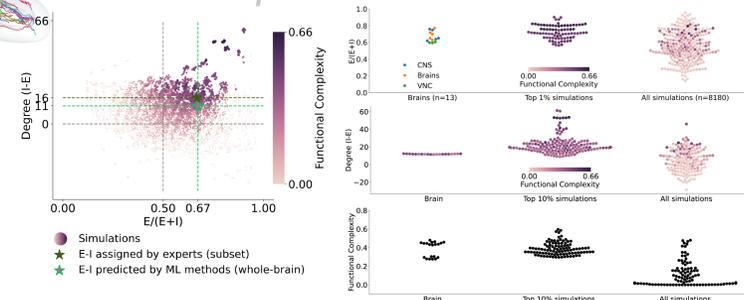
Weight matrix $\mathbf{W} = \alpha_{\text{sign}} \odot \mathbf{M}_{\text{EM}}$

Conditional graph sampling



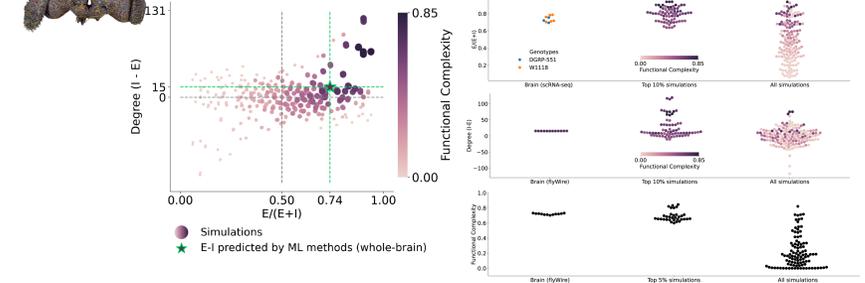
As the inhibitory population is increasingly biased to be highly-connected (darkened colors), the peak functional complexity is reached at higher E/(E+I) ratios, with higher overall functional complexity. $n=10$. Shaded area=95% CI.

Larva Drosophila



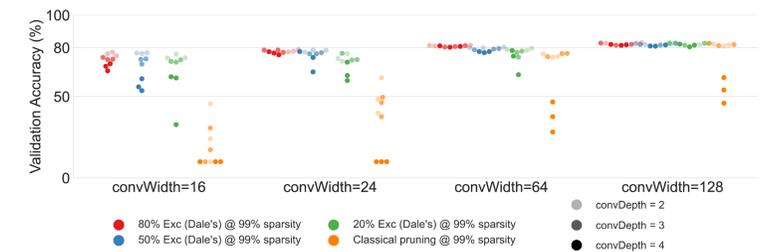
The top simulations of highest functional complexity match the real brain observations in the larva. 8180 different E-I compositions were drawn from 818 different connectedness-conditioned sampling procedures. y-axis = $\text{Degree}_I - \text{Degree}_E$. Networks of higher functional complexity (darker bigger) have both over-abundance of Exc. ($x\text{-axis} > 0.5$) and more highly connected Inhib. compared to the Exc. population ($y\text{-axis} > 0$). They match to brain observations (green stars).

Adult Drosophila



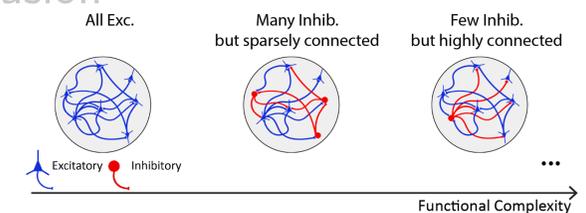
The top simulations of highest functional complexity match the real brain observations in the adult. Same as the larva except 705 different E-I compositions were drawn from 235 different connectedness-conditioned sampling procedures.

Bio-inspired Sparse DNNs



Sparse deep neural networks (DNNs) present a challenge in training (orange); DNNs constrained by Dale's rule and with over-abundance of Exc. (red) solves the challenge the best.

Conclusion



Q: What E-I structure $\Rightarrow \uparrow$ functional complexity?
A: over-abundance of E + highly-connected I.

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