A statistical analysis of bilateral symmetry in an insect connectome

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My requests

- Feedback, feedback, feedback
 - Especially with figures

Electron microscopy connectomics



Winding, Pedigo et al. "The complete connectome of an insect brain." In prep (2021)

Why are we talking about maggot brain?



Drosophila larva (AKA a maggot) brain connectome

See Michael Windings's talk

- First whole-brain, single-cell connectome of any insect
- ~3000 neurons, ~544K synapses
- Both hemispheres of the brain reconstructed



We're just going to consider this to be a network



Why bilateral symmetry?

We examined the connectivity of members of left–right homologous neuron pairs onto left–right homologous targets in the nerve ring of the hermaphrodite reconstruction to assess the amount of natural variability in connectivity. ... Differences between individual worms will be expected to be at least this large. This information is used in the following section to identify sex differences.

In both sexes, the gustatory neuron ASEL (that is, the left neuron of the pair) has greater chemical connectivity than ASER (that is, the right neuron of the pair) to the olfactory neuron class AWC. The ASEL–ASER pair is known to be lateralized in its ability to sense chemosensory cues

Many connectomics questions require comparison

Connectomes across development



Witvliet et al. Nature (2021)

Connectomes across evolution, cortex



So, studying bilateral symmetry here lets us

- Try to formalize what we even mean by this property, and make claims about what we find in this connectome, and
- Test out methods for comparing networks for these future pursuits

Are the left and right sides of this connectome *the same?*

Are these populations the same?

- Known as two-sample testing
- $\bullet~Y^{(1)}\sim F^{(1)}$, $Y^{(2)}\sim F^{(2)}$
- $egin{array}{ll} m{H}_0: m{F}^{(1)} = m{F}^{(2)} \ H_A: m{F}^{(1)}
 eq m{F}^{(2)} \end{array}$



Are these two networks the same?

- Want a two-network-sample test!
- $A^{(L)} \sim F^{(L)}$, $A^{(R)} \sim F^{(R)}$
- $H_0: F^{(L)} = F^{(R)}$ $H_A: F^{(L)} \neq F^{(R)}$



Assumptions

- We know the direction of synapses, so network is *directed*.
- For simplicity (for now), consider networks to be *unweighted*.
- For simplicity (for now), consider the left → left and right → right (ipsilateral) connections only.
- Not going to assume any nodes are matched



Density-based testing: Erdos-Renyi (ER) model



We detect a difference in density



• p-value < 10^{-22}

Group-based testing: stochastic block model (SBM)



Connection probabilities between groups







We detect a difference in group-to-group connection probabilities



- After multiple comparison, find 5 group-to-group connections which are significantly different
- Combine (uncorrected) p-values (like a meta-analysis), leads to p-value for overall test of $< 10^{-7}$

A massive aside

Combining p-values: nobody's perfect



Combining p-values: don't trust scipy

elif method == 'tippett':

statistic = np.min(pvalues)

- pval = distributions.beta.sf(statistic, 1, len(pvalues))
- pval = distributions.beta.cdf(statistic, 1, len(pvalues))

Distribution under the null for combining pvalues



Combining p-values: be careful with discreetness

THE COMBINATION OF PROBABILITIES ARISING FROM DATA IN DISCRETE DISTRIBUTIONS

BY H. O. LANCASTER, Rockefeller Fellow in Medicine



 \leftarrow We are trying to approximate this null distribution with something continuous Uniform(0,1)

Power for combining pvalues

- We perturb:
 - Some # of them (x-axis)
 - By some amount (panels)



Relative power (Fisher's vs Tippett's)



Back to the main thread...

Should we be surprised?

- Already saw that even the overall densities were different
- For all significant comparisons, probabilities on the right hemisphere were higher
- Maybe the right is just a "scaled up" version of the left?

 $\circ \ H_0: B^{(L)} = c B^{(R)}$ where c is a density-adjusting constant, $rac{p^{(L)}}{p^{(R)}}$



Adjusting for a difference in density



Even with density adjustment, we detect a difference



So the Kenyon cells are the only group with remaining differences...











aSBM

H_0 :	$B^{(L)}$	= <i>c</i>	B ^(R)
H_A :	$B^{(L)}$	≠ C	B ^(R)

- ER test: $p < 10^{-26}$
- SBM test: ppprox 0.0027
- Adjusted SBM test: ppprox 0.43

To sum up...

Model	H_0 (vs. $H_A eq$)	KC	p-value	Interpretation
ER	$p^{(L)} = p^{(R)}$	+	$< 10^{-23}$	Reject densities the same
SBM	$B^{(L)} = B^{(R)}$	+	$< 10^{-7}$	Reject group connection probabilities the same
aSBM	$B^{(L)} = cB^{(R)}$	+	pprox 0.0016	Reject above even after accounting for density
ER	$p^{(L)} = p^{(R)}$	-	$< 10^{-26}$	Reject densities the same (w/o KCs)
SBM	$B^{(L)} = B^{(R)}$	-	pprox 0.0027	Reject group connection probabilities the same (w/o KCs)
aSBM	$B^{(L)} = cB^{(R)}$	-	pprox 0.43	Don't reject above after density adjustment (w/o KCs)

Extensions

But you threw out all of the edge weights!



There are so many other models!

Latent distribution test (random dot product graph)



In summary...

- We studied simple ways of framing a network two sample test, and proposed test procedures for each
 - We found that it can be important to "mod out" by other simple network statistics if you don't care about them (like density)
- We found that all of these tests find the left and the right hemispheres to be significantly different, unless you ignore Kenyon cells and adjust for the difference in density
- The tests proposed here provide a foundation for future principled comparisons of connectomes

graspologic:

github.com/microsoft/graspologic





This work:

github.com/neurodata/bilateralconnectome





The Erdos-Renyi (ER) model

The Erdos-Renyi (ER) model is one of the simplest network models. This model treats the probability of each potential edge in the network occuring to be the same. In other words, all edges between any two nodes are equally likely.

Let n be the number of nodes. We say that for all $(i, j), i \neq j$, with i and j both running from $1 \dots n$, the probability of the edge (i, j) occuring is:

 $P[A_{ij} = 1] = p_{ij} = p$

Where p is the the global connection probability.

Each element of the adjacency matrix A is then

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Questions?

Left

Right