

The Neural Processes of Statistical Learning

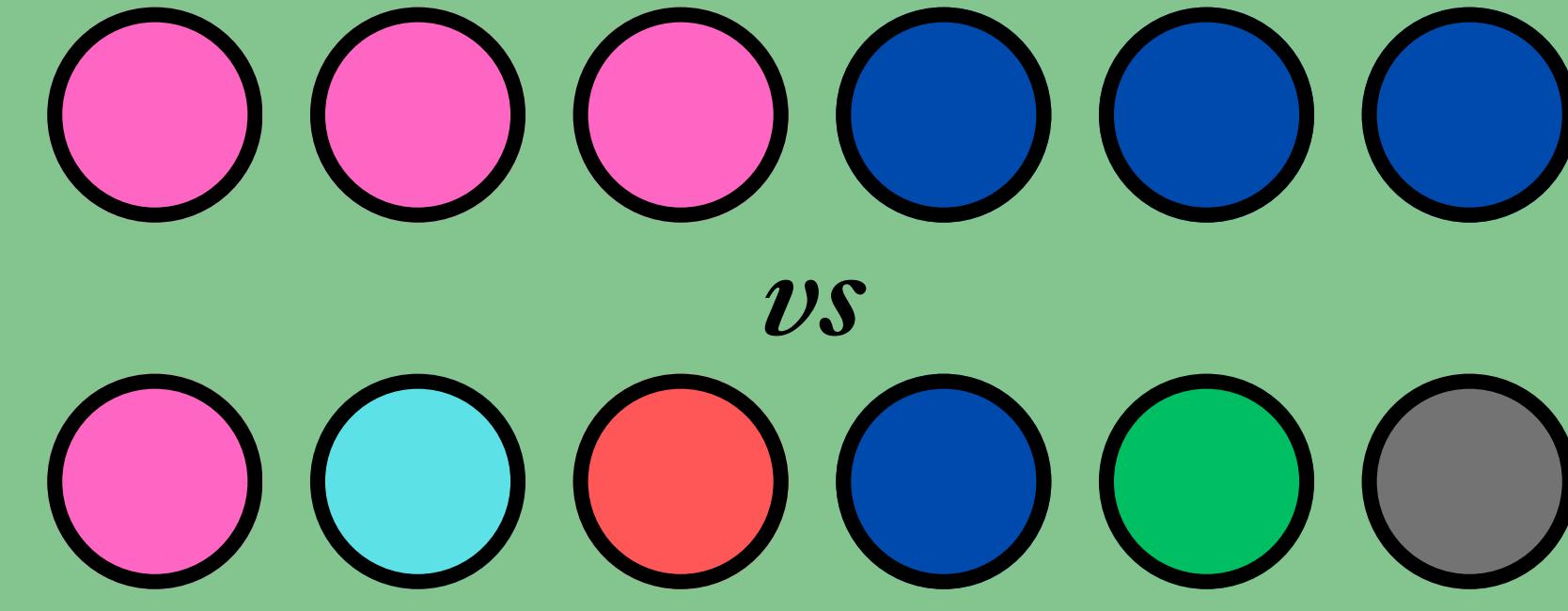
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Background

What is statistical learning?



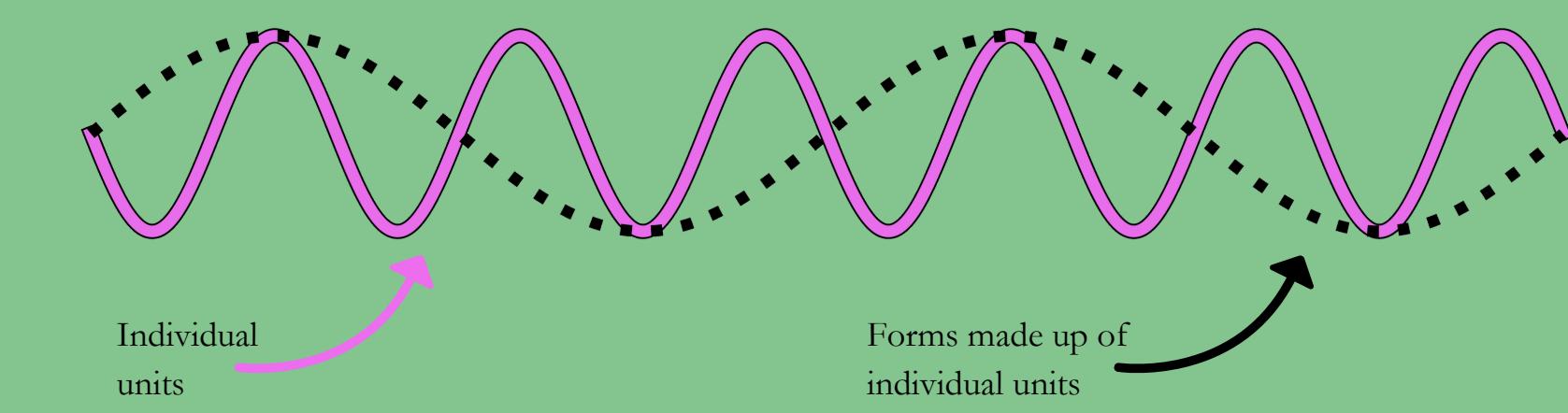
- Statistical learning is the ability to extract units of information based on patterns of regularities in perceptual input.
- Theory implicates statistical learning as a critical mechanism for language acquisition.¹¹
- Recent works suggest statistical learning is not a unitary process, but occurs along varying spatial and temporal scales in the brain.⁵

Computational formalization

$$P(Y|X) = \frac{P(X \cap Y)}{P(X)}$$

Likelyhood stimulus **Y** will co-occur with **X** given the frequency of stimulus **X**.⁶

Neural entrainment

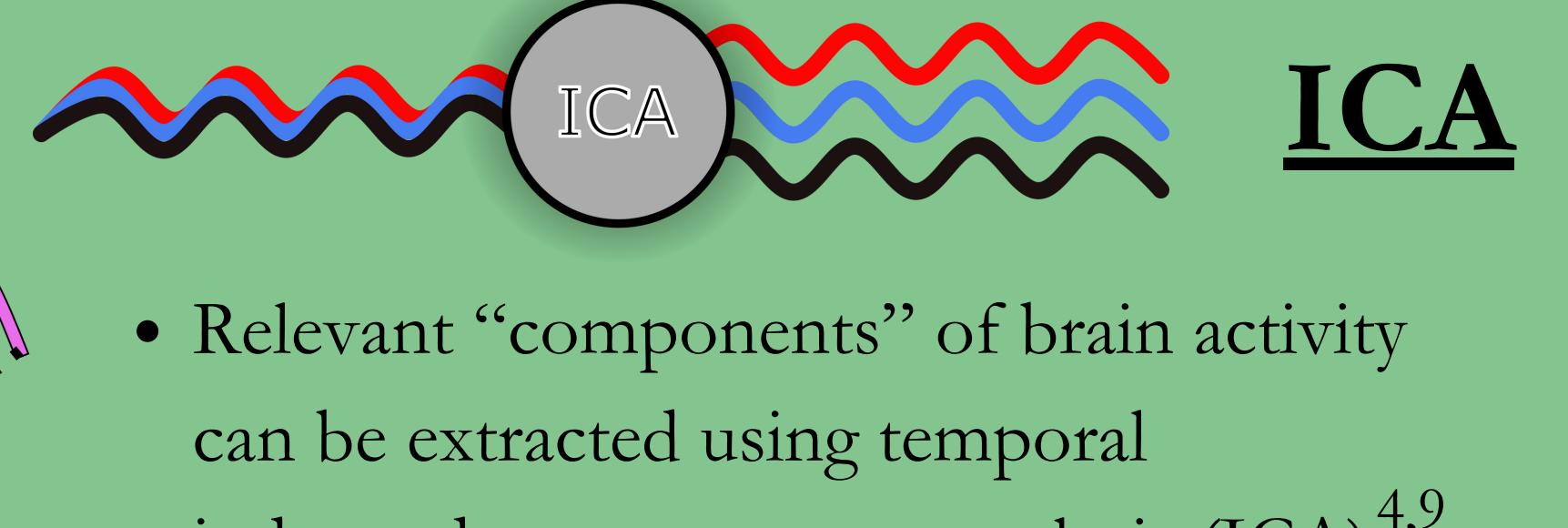


- Brain activity oscillates at the same frequency of rhythmic stimuli as a function of statistical learning.²
- The methods used to detect neural entrainment (e.g., M/EEG) have low spatial resolution.

Linguistic example

Cool dads

Syllables *within* words have higher transitional probabilities than syllables *across* words.



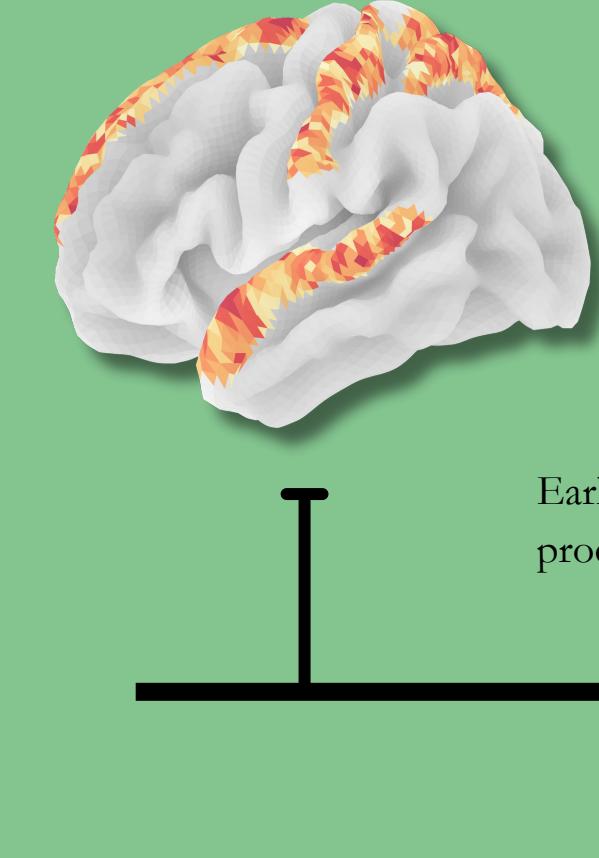
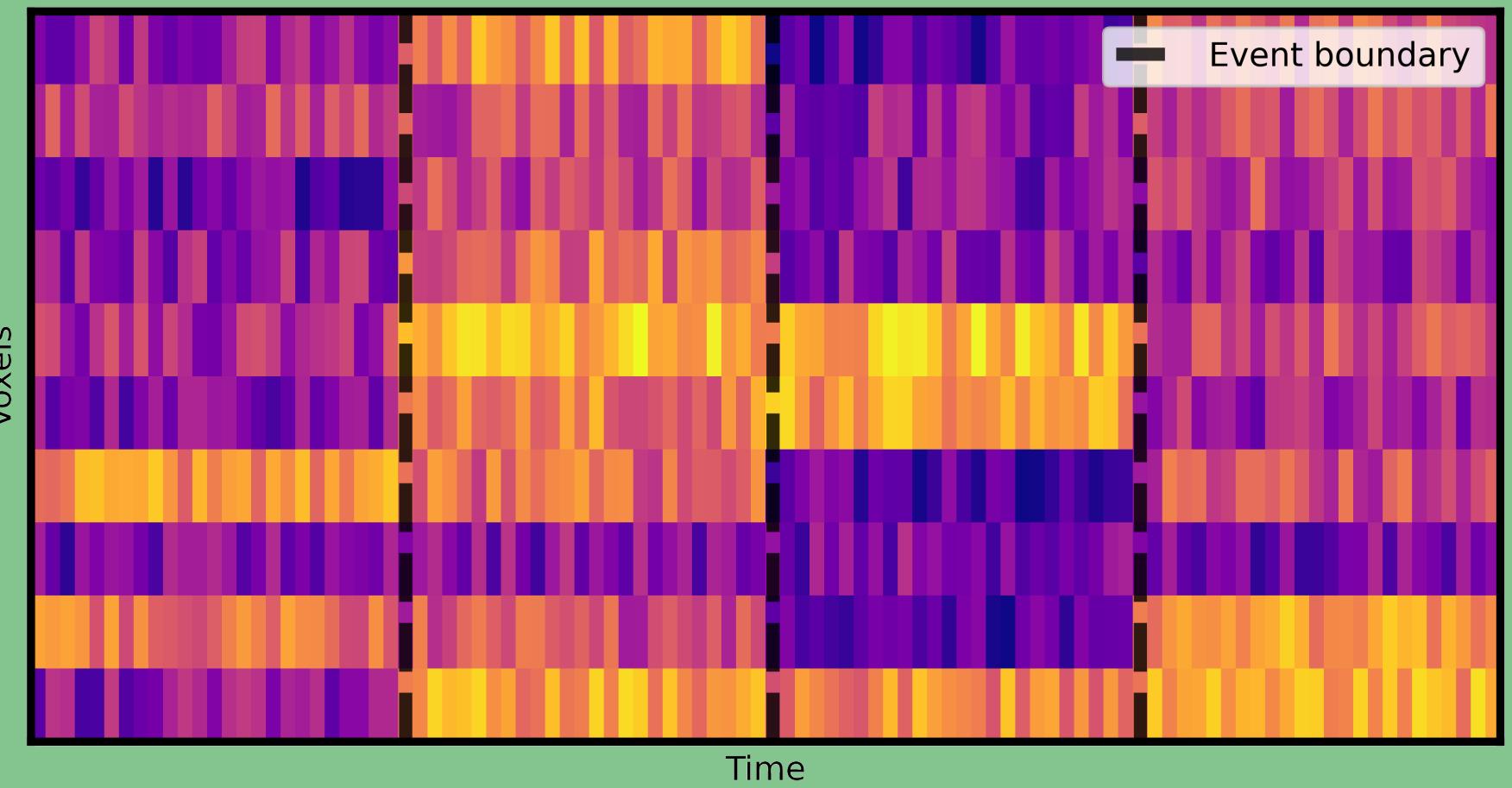
- Relevant “components” of brain activity can be extracted using temporal independent component analysis (ICA).^{4,9}

Can we comprehensively characterize statistical learning in the brain as it unfolds over space and time with fMRI?

Methods

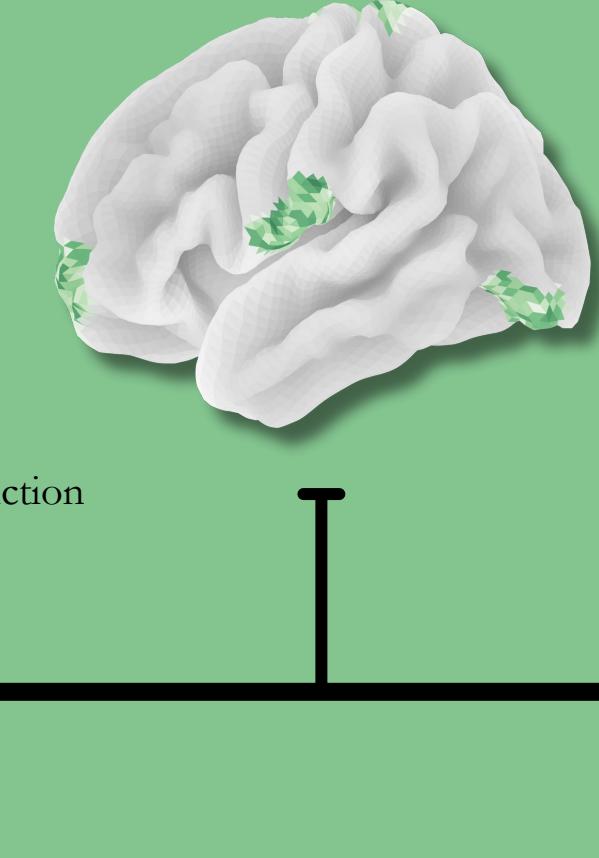
Event Segmentation

- Brain activity is *not random*. It can exist in various states.^{1,14}
- Can we uncover distinct event-states of *statistical learning*?



Dynamic Functional Connectivity

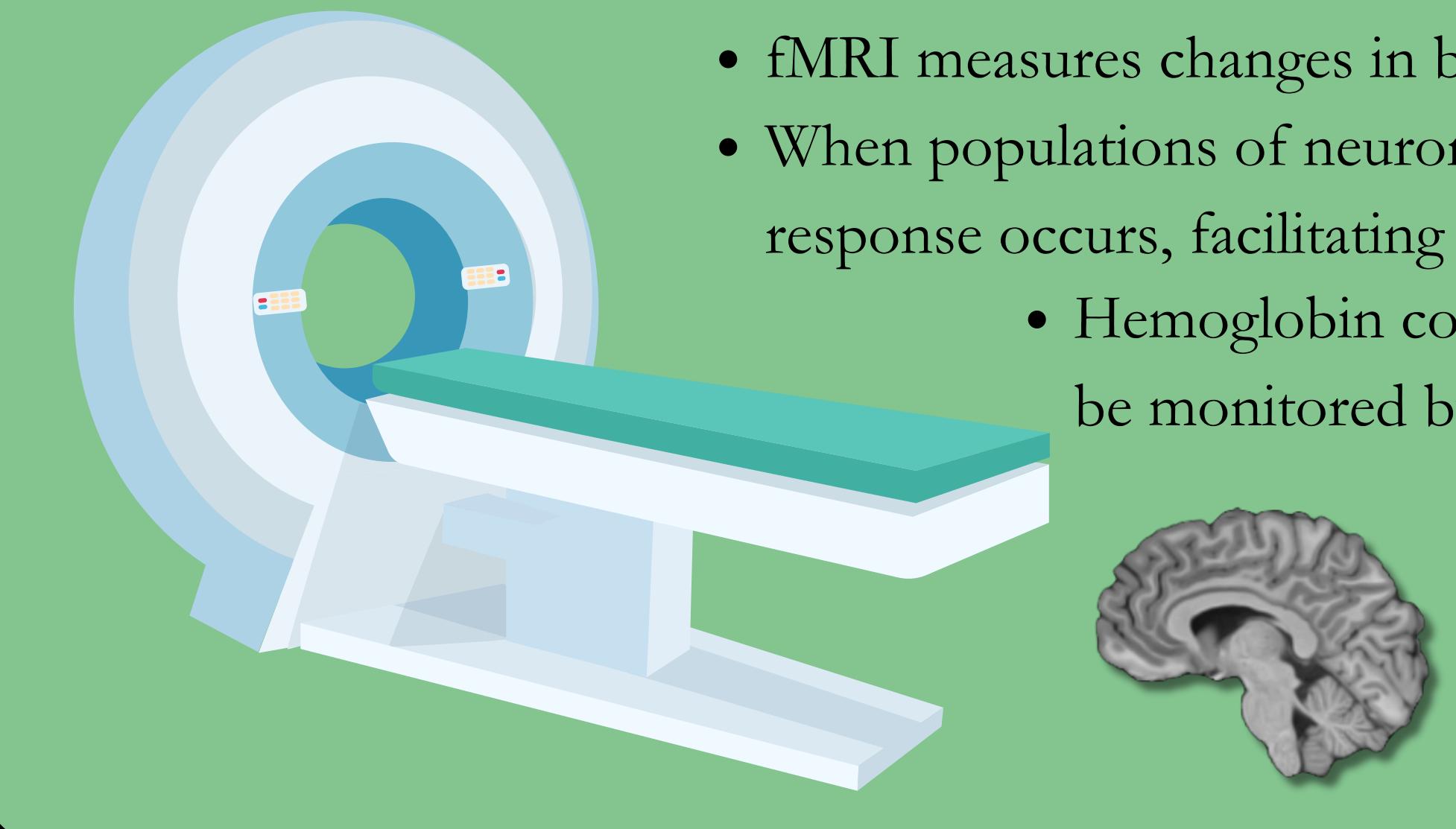
- Brain regions are “functionally” connected; their activity is highly correlated despite being spatially distinct.³
- Can we reveal changes in functional connectivity as a consequence of *statistical learning*?



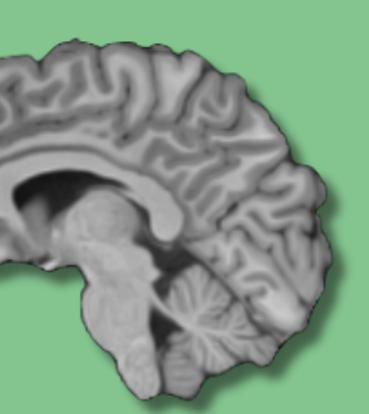
Changes in correlated brain activity during *statistical learning*.

Are their systematic profiles of correlated neural activity at different times during learning?

Functional Magnetic Resonance Imaging



- fMRI measures changes in blood flow inside the brain.
- When populations of neurons are firing, a haemodynamic response occurs, facilitating neural activity.
- Hemoglobin contains magnetic properties which can be monitored by the fMRI scanner.



In-scanner statistical learning task



Out-of-scanner behavioral test of learning



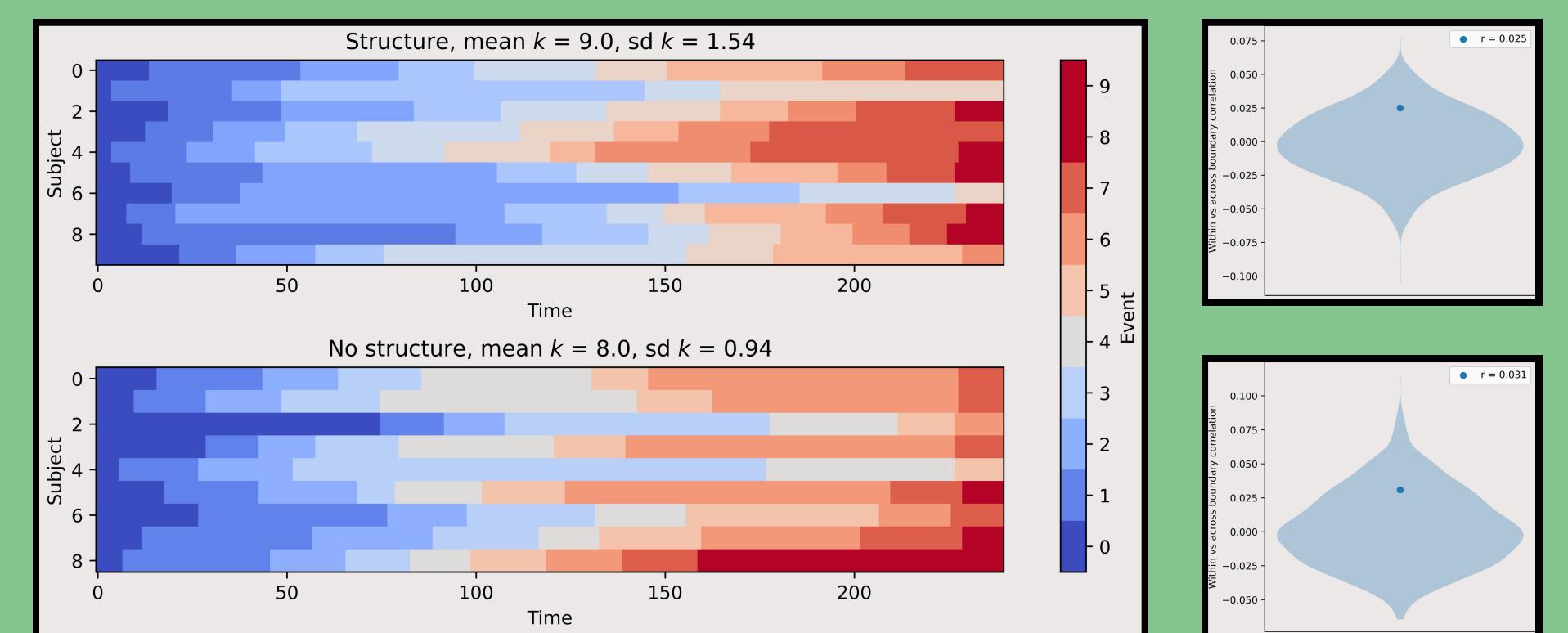
Data cleaning, formal analyses, statistical evaluations

Procedures

Results & Discussion

Event Segmentation

- An adapted hidden Markov model (HMM) was used to calculate event structure from the continuous fMRI time series.^{1,8}

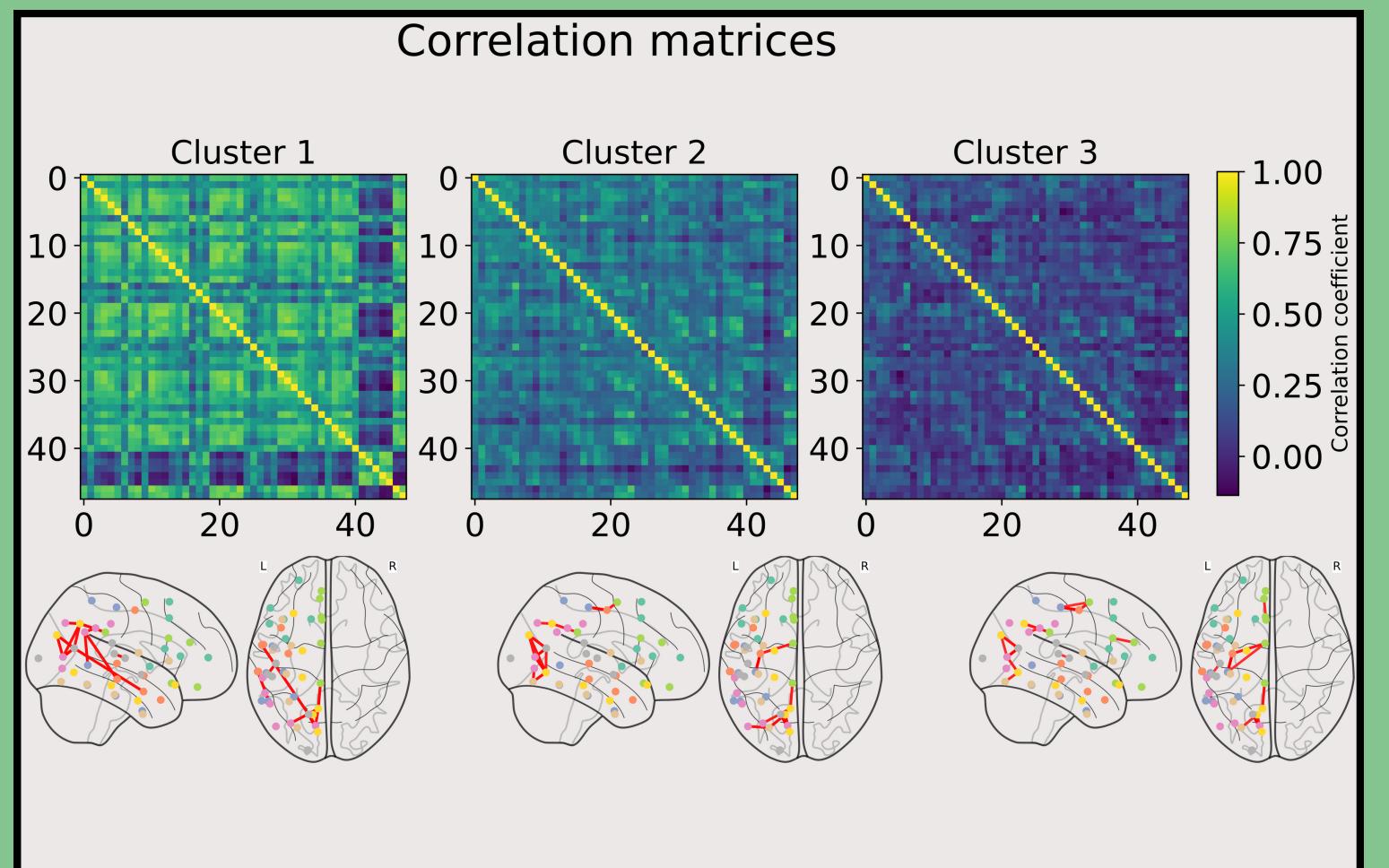


- Although behavioral results confirmed that learning occurred, subjects’ neural data is more variable when learning than when not. This could be due to dimensionality reduction approach. Or....
- These group differences demonstrate the heterogeneity of learning: brain data is more variable when learning compared to not learning.^{7,12}

Behavior

Dynamic Functional Connectivity

- Using a sliding window approach, we calculated dynamic functional connectivity for the structured group using Pearson’s r^{13}
- Presence of low connectivity “state” inline with previous work.¹⁰
- Cluster one shows particular emphasis on auditory cortex.
- Left lateralization and general patterns bolster the idea that domain-specific perception and higher-order processing scaffold *statistical learning*. This is grounded in theoretical work.^{5,7}



Bib

- Baldassarre, C., Chervi, J., Zadra, A., Silver, J. W., Harvey, U., & Norman, K. A. (2017). Decoding event structure in communication perception and memory. *NeuroImage*, 95, 70–72. <https://doi.org/10.1016/j.neuroimage.2017.06.041>
- Bonin, J. L., & Price, C. A. (2017). Online neural monitoring of statistical learning. *Cortex*, 53, 31–43. <https://doi.org/10.1002/cort.22460>
- Chen, X., & Li, H. (2017). fMRI-based brain activity monitoring of statistical learning. *Frontiers in Neuroscience*, 9, 1–10. <https://doi.org/10.3389/fnins.2017.00001>
- Gholson, V. D., Addis, T., Pfeffer, G. D., & Polka, J. J. (2001). A method for making group inferences from functional MRI data using independent component analysis. *Human Brain Mapping*, 14(3), 140–150. <https://doi.org/10.1002/hbm.10014>
- Gruber, C. M. (2013). How does the brain learn statistical structure? Two core principles for understanding the neurocognitive mechanisms of statistical learning. *Neuroscience & Biobehavioral Reviews*, 112, 279–299. <https://doi.org/10.1016/j.neurobioreviews.2013.03.002>
- Gruber, S. N., & Gruber, C. M. (2017). Unpacking Vowel Probability Differences Between Theories of Statistical Learning. *Cognition*, 155, 132–144. <https://doi.org/10.1016/j.cognition.2017.04.012>
- Troyer, B., & Gruber, C. M. (2017). Statistical learning and its neural correlates. *Neuroscience & Biobehavioral Reviews*, 112, 250–278. <https://doi.org/10.1016/j.neurobioreviews.2017.03.002>
- Kosner, M., Anderson, M. J., Arnon, J. W., Baldassarre, C., Beucke, J. P., Cao, M. R., Cox, P. C., Ellis, C. T., Hammarlund, R., Hoberman, J. R., Ilia, Q., Manning, J. R., Mentzen, S. A., Rehfeld, H., Schepens, A., Schindler, J., Shulman, G., & Norman, K. A. (2017). fMRI monitoring of statistical learning. *Cortex*, 53, 34–42. <https://doi.org/10.1002/cort.22461>
- López-Burillo, D., Argandoña, P., Martí-Peláez, J., Mohamed, R., Monte, J. F., Pachón-Ley, A. C., Rodríguez-Fernández, A., & De Dreu-Balaguer, R. (2013). Multiple brain networks underlying word learning from fluent speech revealed by independent component analysis. *Journal of Cognitive Neuroscience*, 25(10), 1661–1674. https://doi.org/10.1162/jcn_a_00482
- Poldrack, J., Jancke, K., & Soller, J. B. (2010). Reduced functional connectivity in the brain during statistical learning. *NeuroImage*, 50, 396–404. <https://doi.org/10.1016/j.neuroimage.2009.10.049>
- Roskam, A. B., & Soller, J. B. (2010). Statistical learning and language acquisition: Why interdisciplinary research is needed. *Language, Cognition and Neuroscience*, 25(10), 1643–1660. <https://doi.org/10.1080/20544108.2010.497024>
- Scarborough, S. P., & Dobrich, W. (2000). Individual differences in children's word learning: A developmental study. *Memory & Cognition*, 28(1), 1–13. <https://doi.org/10.3758/bf03192083>
- Seidenberg, S. C., McClelland, J. L., Patterson, K., & Bishop, D. V. M. (2003). The variability of dynamic functional connectivity assessment methods (p. 2032). <https://doi.org/10.1162/089826003764562083>
- Strobl, G., Mönig, C. D., & Polka, J. J. (2012). On the variability of dynamic functional connectivity assessment methods. *bioRxiv*. <https://doi.org/10.1101/133883>
- Van Esch, J., Mlynář, N., Kralovič, K., Beucke, J. S., Rehfeld, H., & De Dreu-Balaguer, R. (2017). fMRI brain responses. *Psychological Bulletin*, 152(2), 274–293. <https://doi.org/10.1037/bul0000125>

Ack.

Thank you to the LEND Lab and ReLearn Lab for their continued support.

Some brain plots were generated with my open source Python module, **roipoly**. Check it out!
 * `pip install roipoly`

All code for formal analyses is hosted on GitHub. Scan the QR code.

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