

# The Neural Processes of Statistical Learning

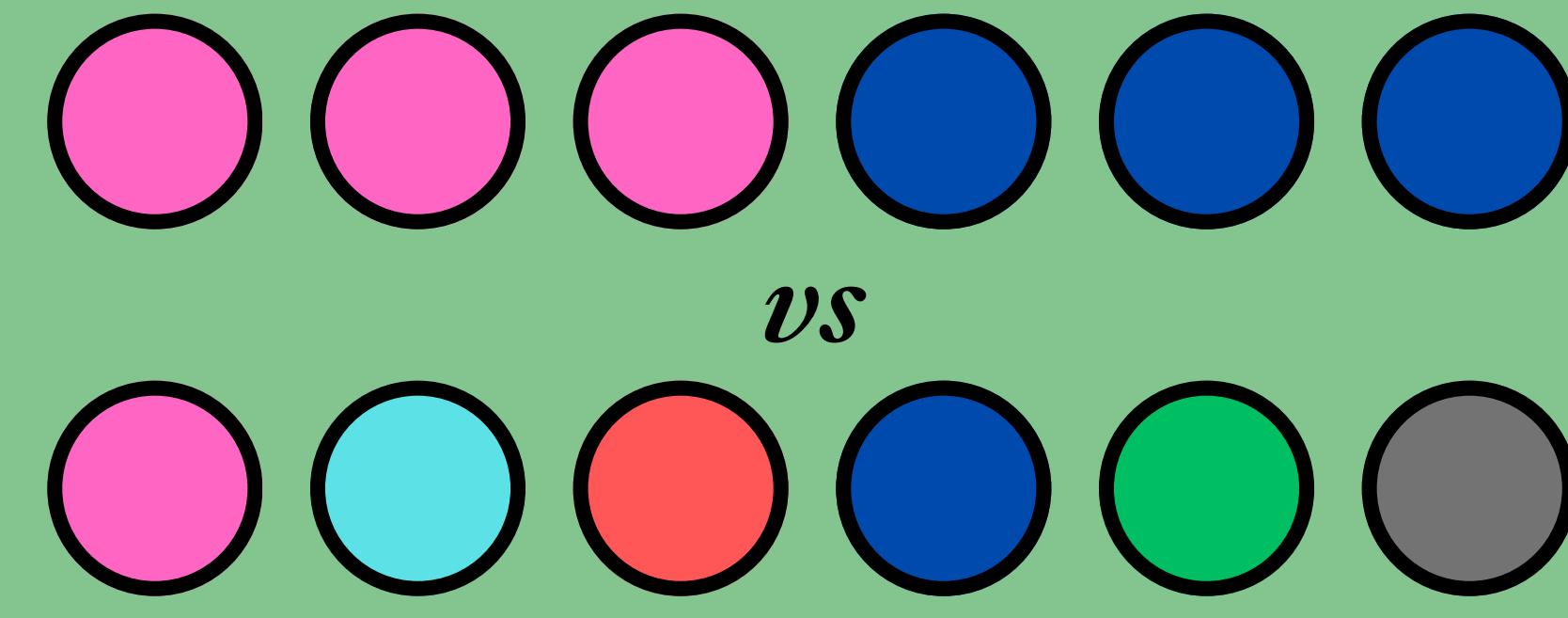
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## Background

### What is statistical learning?



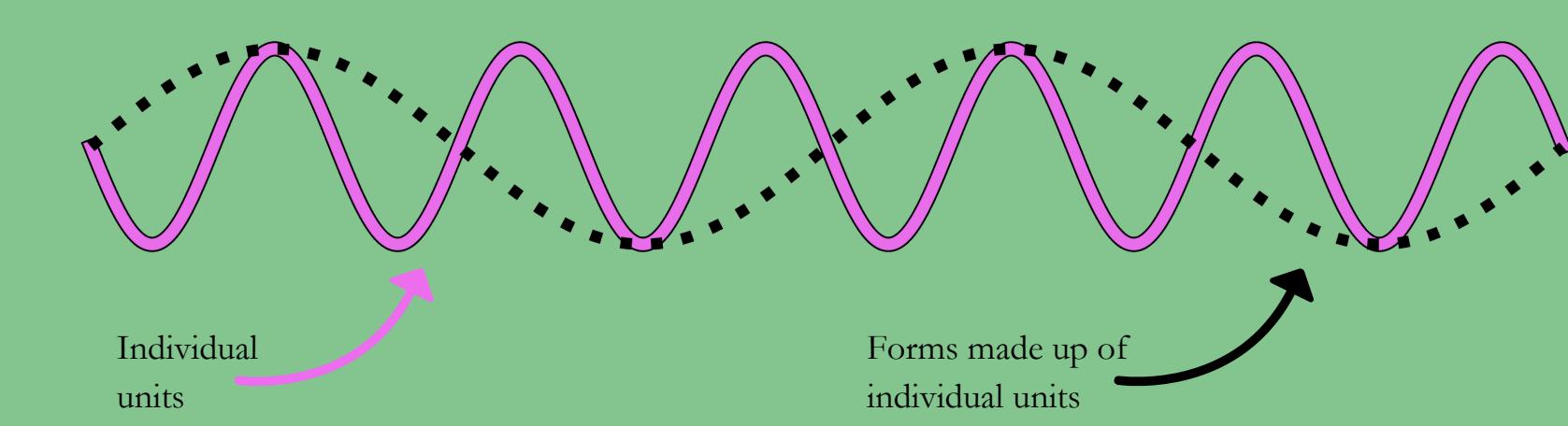
- Statistical learning is the ability to extract units of information based on statistical regularities of the environment.
- Theory implicates statistical learning as a critical mechanism for language acquisition.<sup>11</sup>
- Recent works suggest statistical learning is not a unitary process, but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

### 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**.<sup>6</sup>

### Neural entrainment



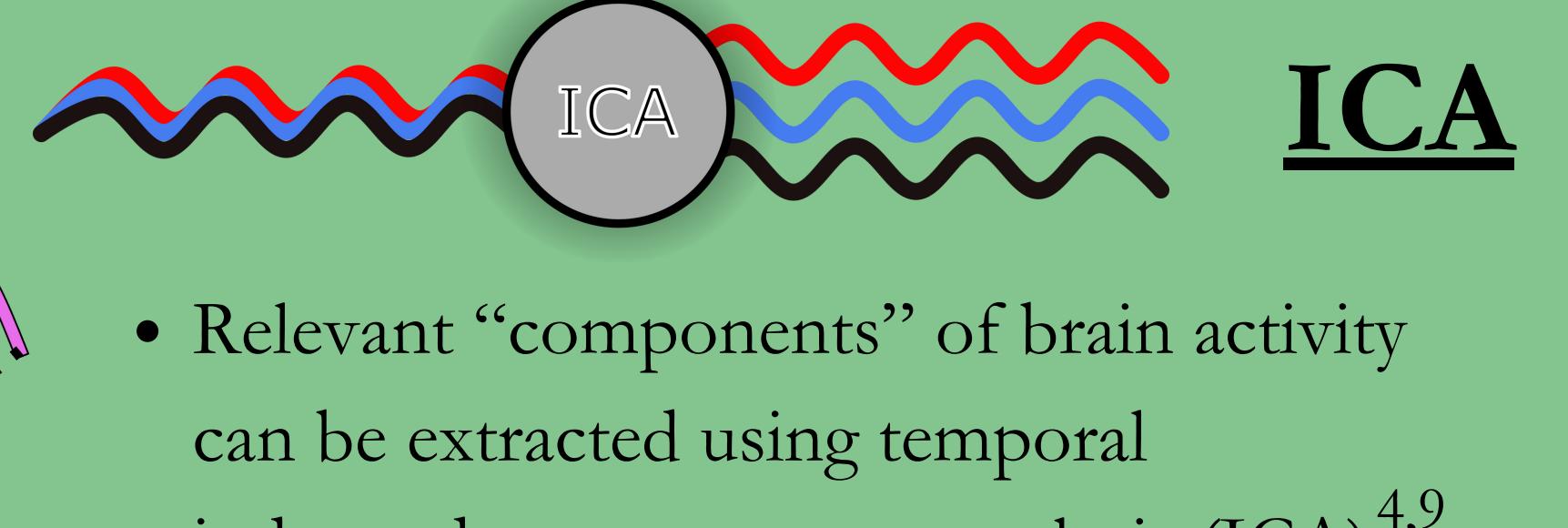
- Brain activity oscillates at the same frequency of rhythmic stimuli as a function of statistical learning.<sup>2</sup>
- The methods used to detect neural entrainment (e.g., M/EEG) have low spatial resolution.

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

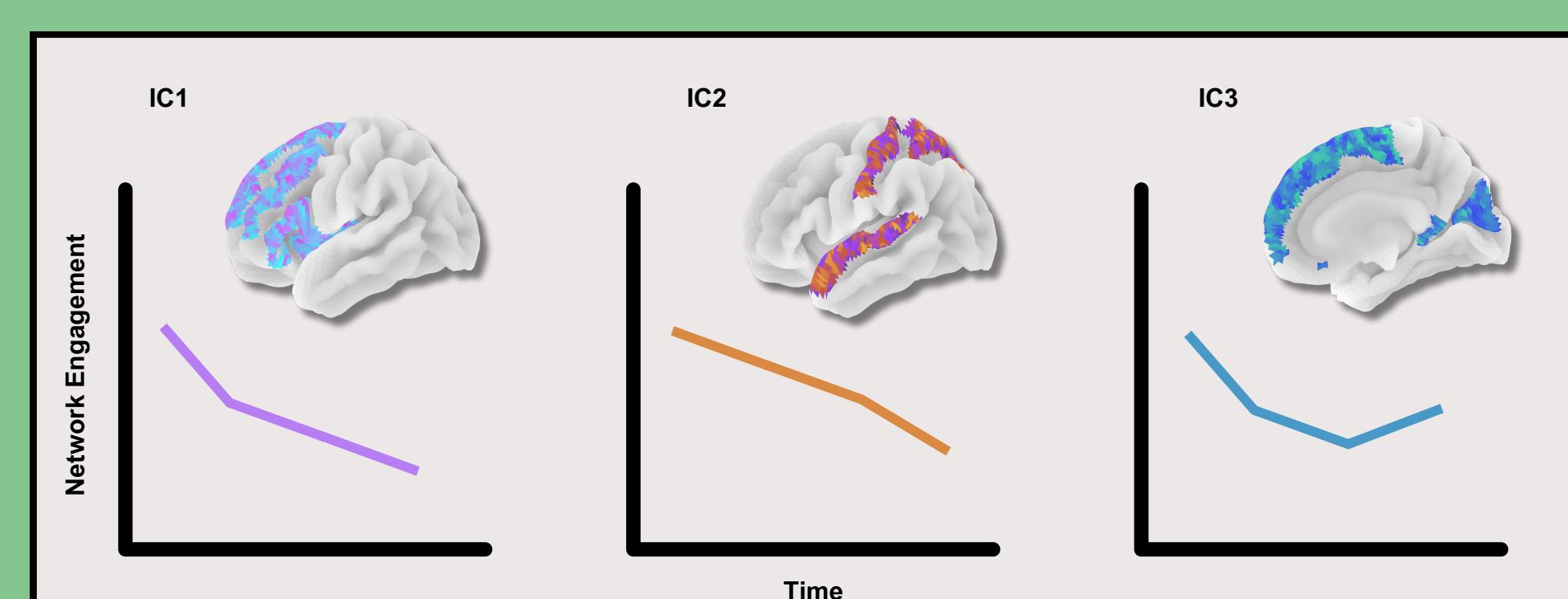
### Linguistic example

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).<sup>4,9</sup>

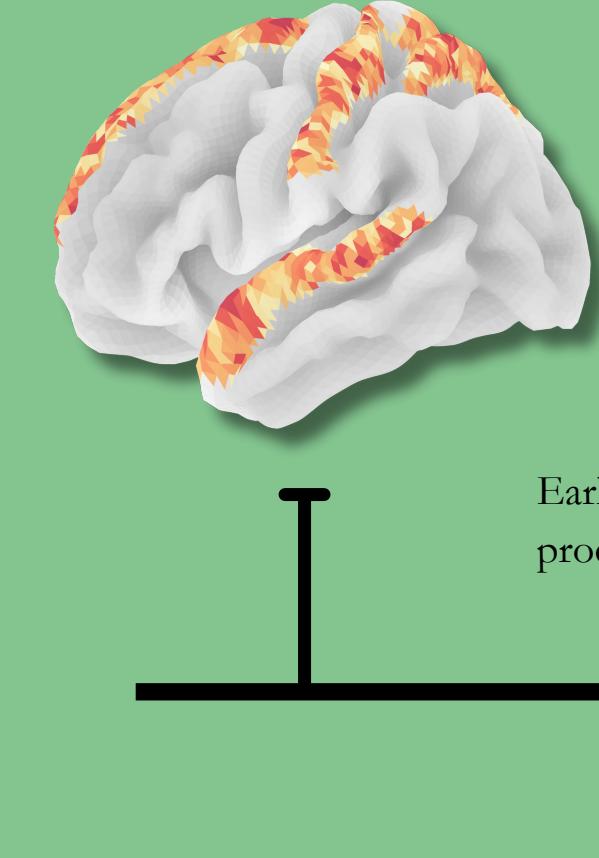
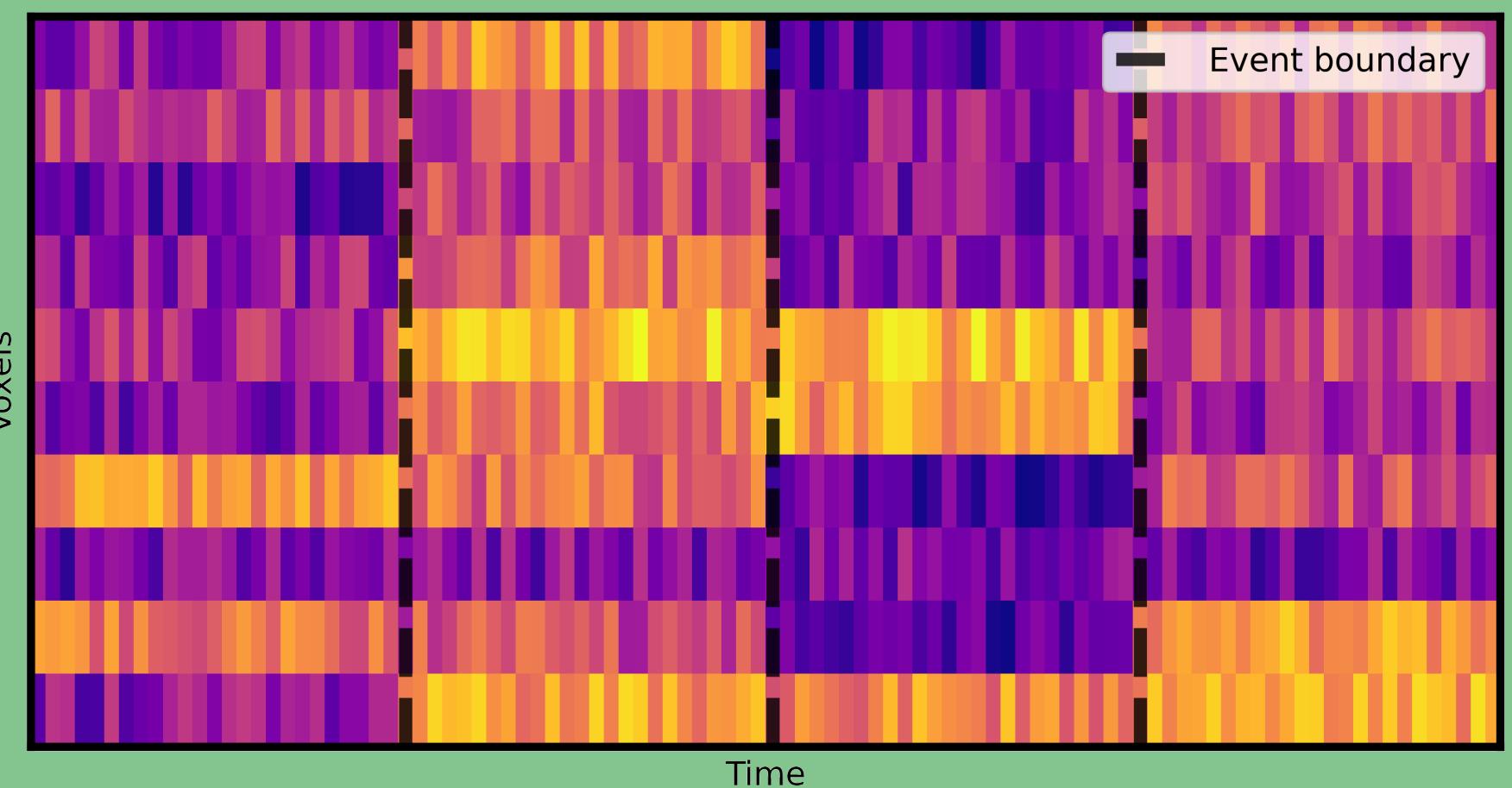


## Event Segmentation

## Methods

### Event Segmentation

- Brain activity is *not random*. It can exist in various states.<sup>1,14</sup>
- Can we uncover different event-states of *statistical learning*?



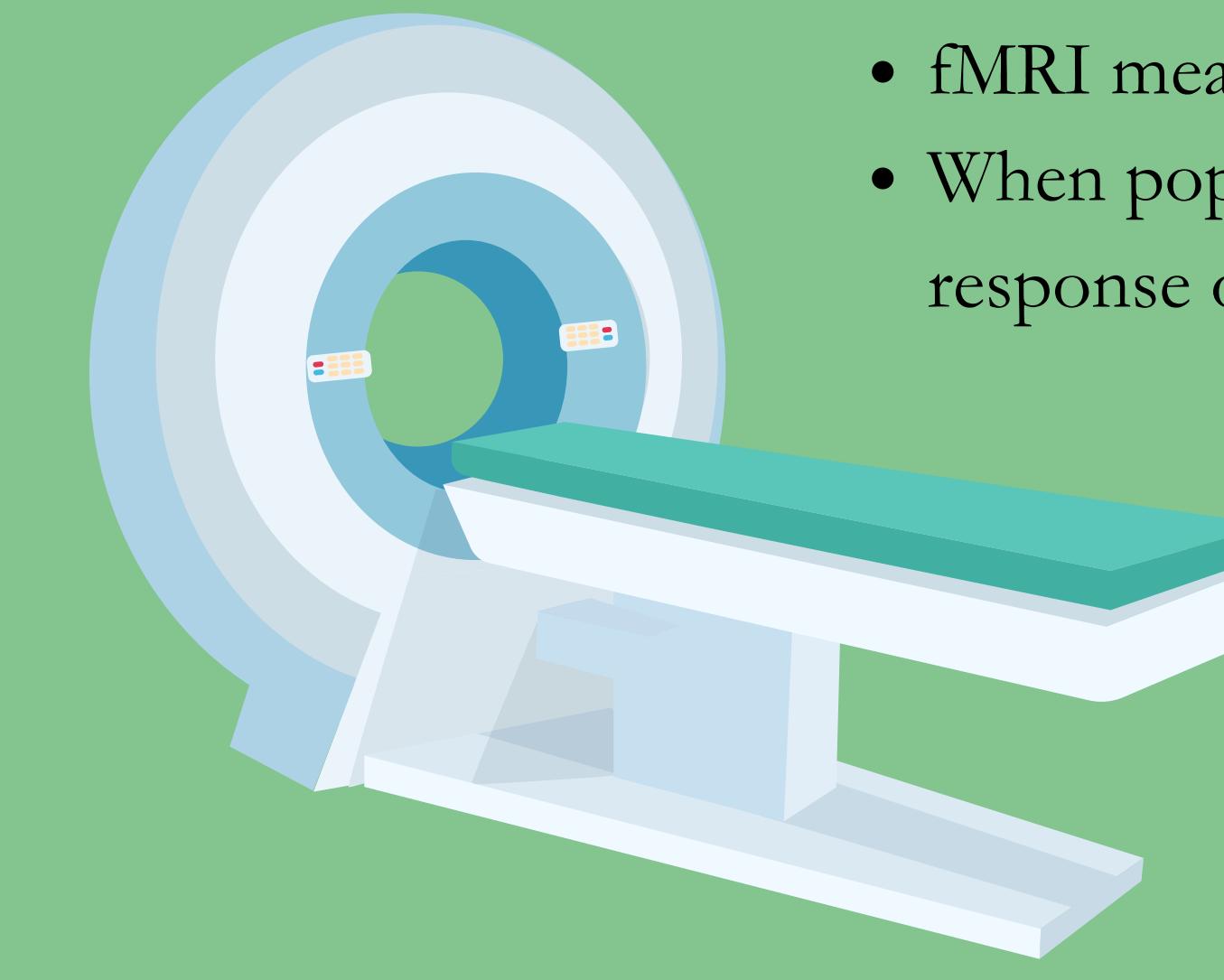
### Dynamic Functional Connectivity

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

Changes in correlated brain activity during *statistical learning*.



### Functional Magnetic Resonance Imaging



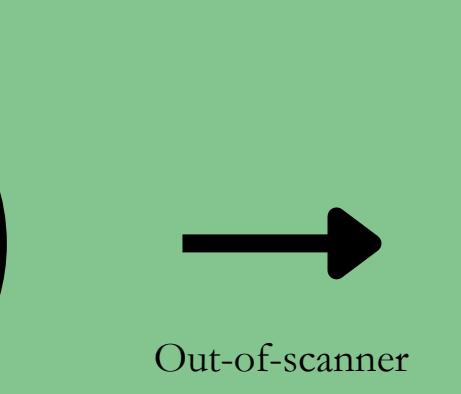
- fMRI measures changes in blood flow inside the brain.
- When populations of neurons are firing, a haemodynamic response occurs, facilitating neural activity.
- Hemoglobin contains magnetic properties which can be monitored in the fMRI scanner.



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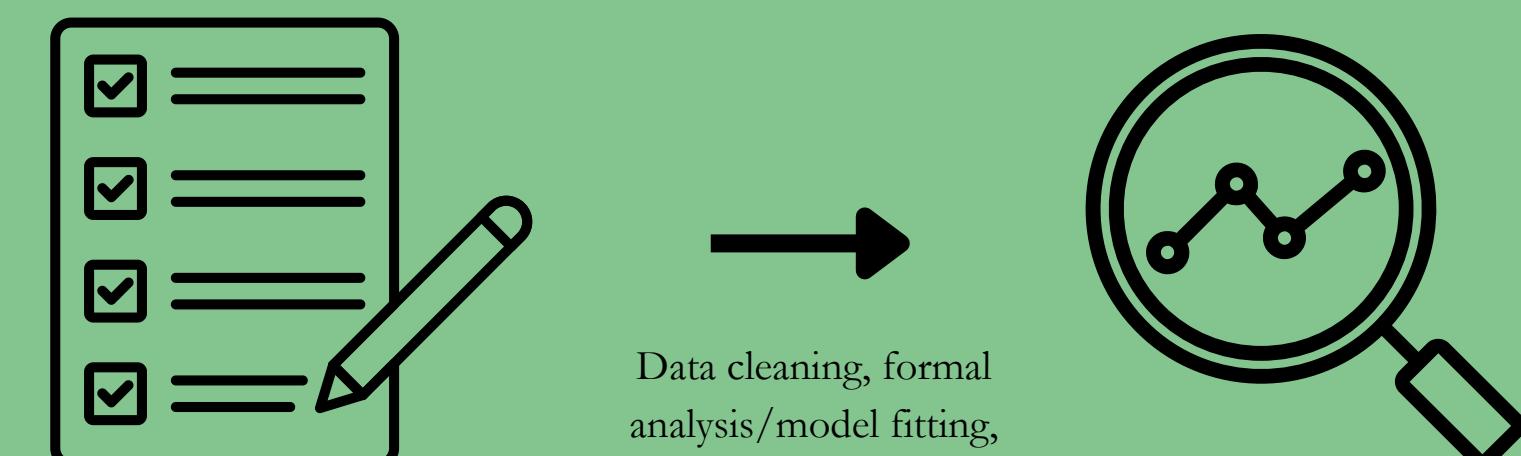
In-scanner statistical learning task



Out-of-scanner behavioral test of learning

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

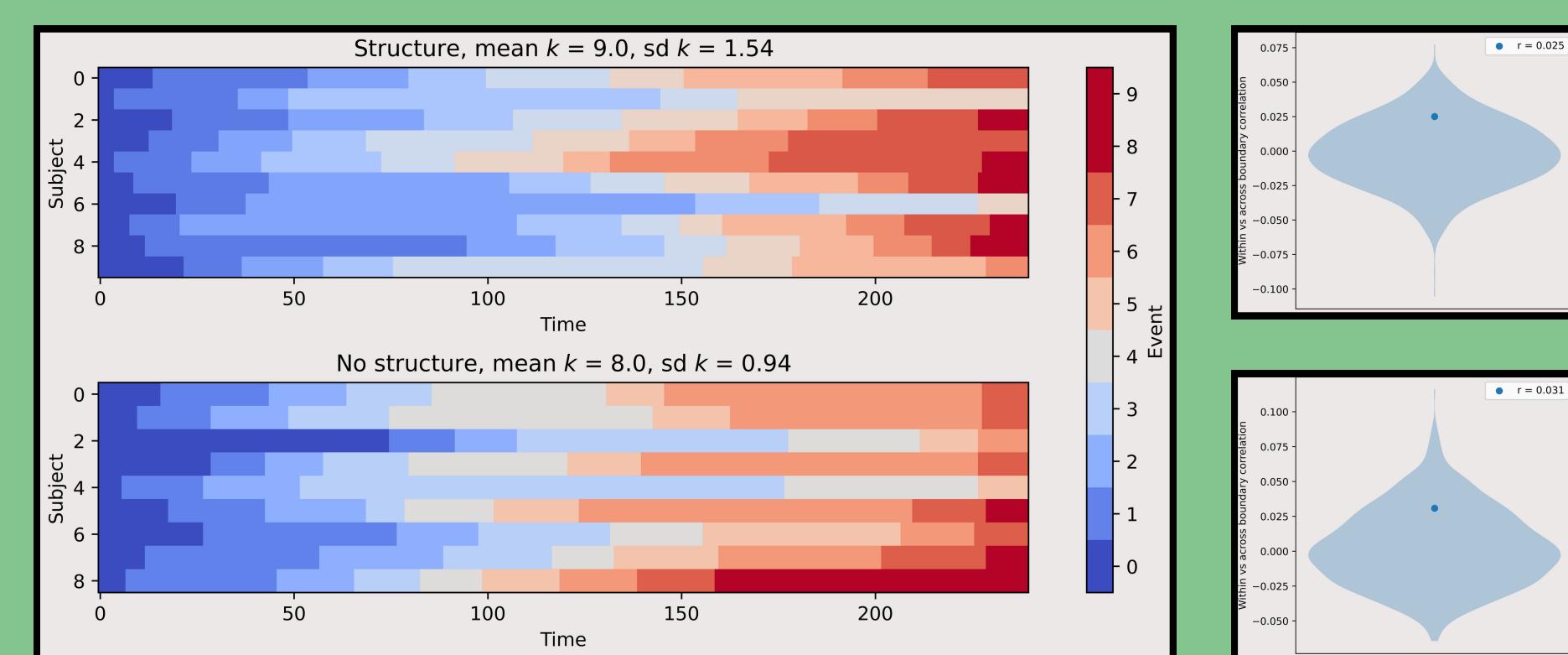
### Procedures



## Results & Discussion

### Event Segmentation

- An adapted Hidden Markov Model (HMM) was used to calculate event structure from the continuous fMRI timeseries.<sup>1,8</sup>

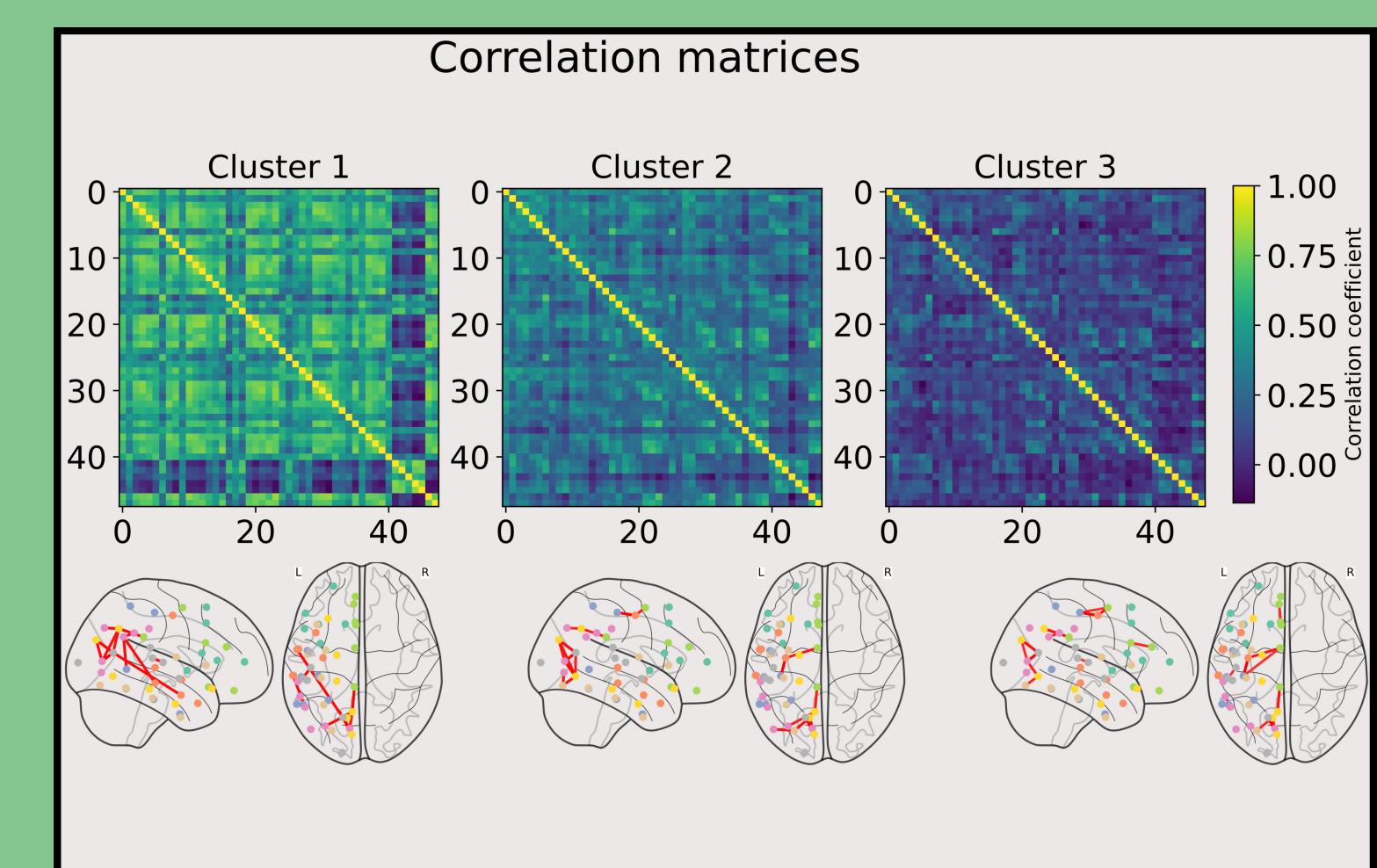


- Although behavioral results confirmed that learning occurred, subjects’ neural data is more variable when learning than when not. This could be due to dimensionality reduction approach. Or....
- These group differences demonstrate the heterogeneity of learning: brain data is more variable when learning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

### Behavior

### Dynamic Functional Connectivity

- Using a sliding window approach, we calculated dynamic functional connectivity for the structured group using Pearson’s  $r^{13}$
- Overall reduction in FC is inline with previous work.<sup>10</sup>
- 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.<sup>5,7</sup>



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All code for formal analyses is hosted on GitHub. Scan the QR code.  
\* `pip install roipy`  
Check it out!

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