

# Cognitive correlates of BOLD Resting-State Dynamic Functional Connectivity

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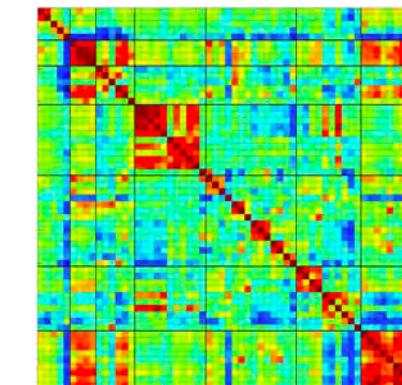
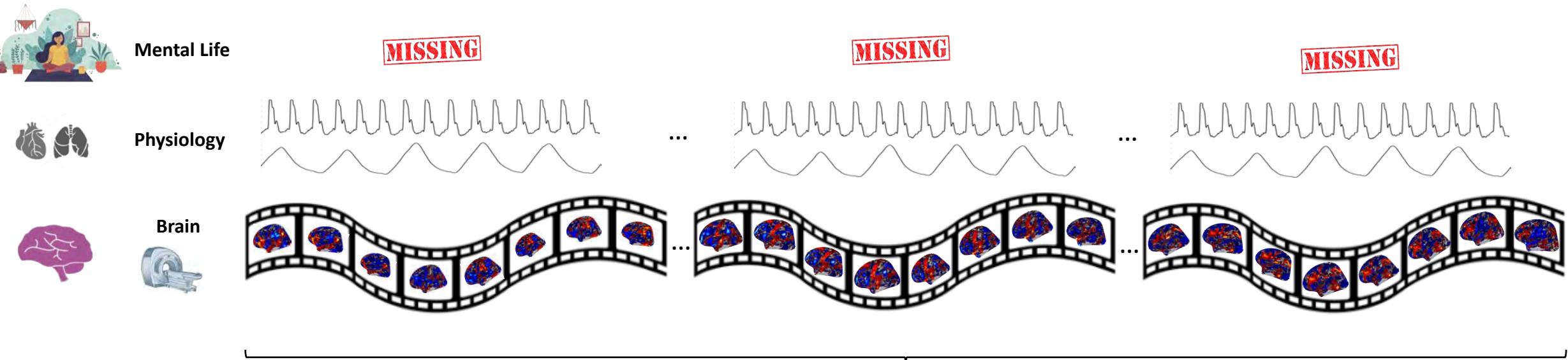
Section on Functional  
Imaging Methods



National Institute  
of Mental Health



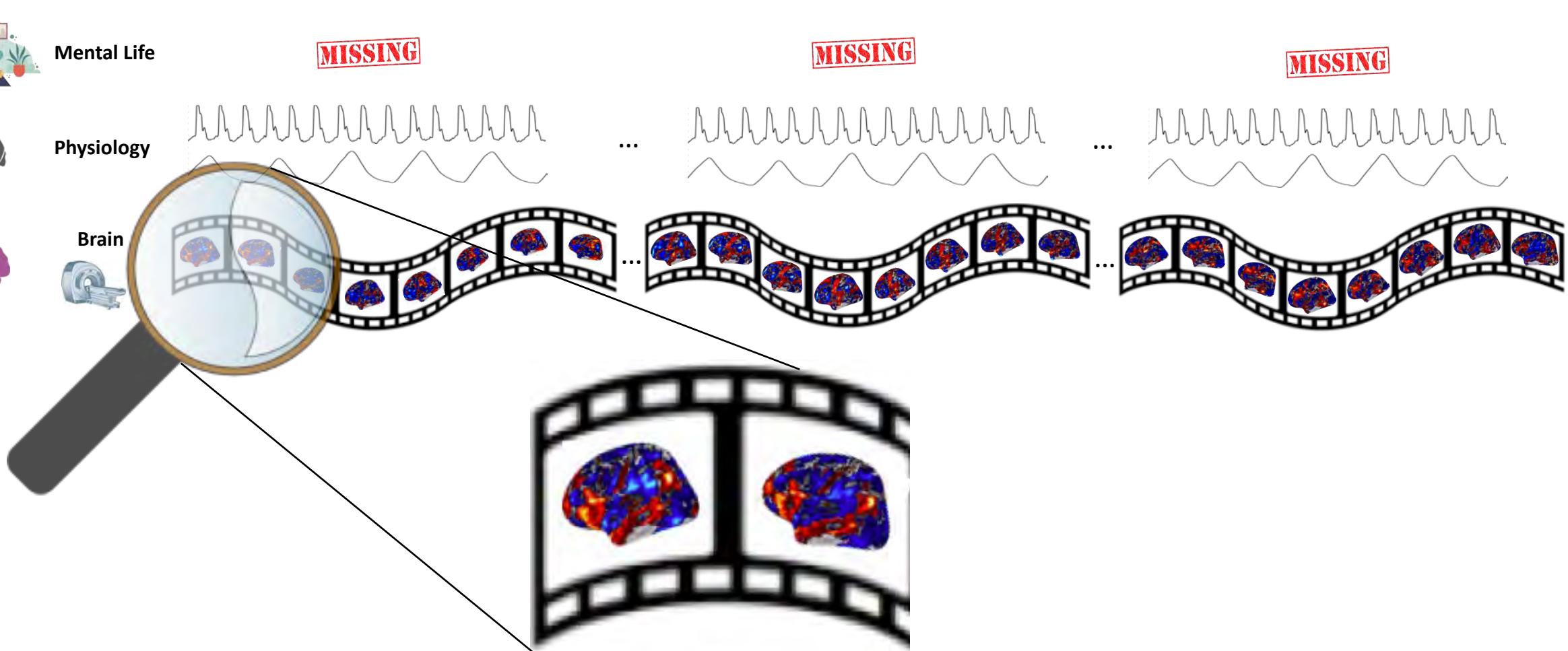
# Introduction: Resting State



Mean Effect – Static Functional Connectivity

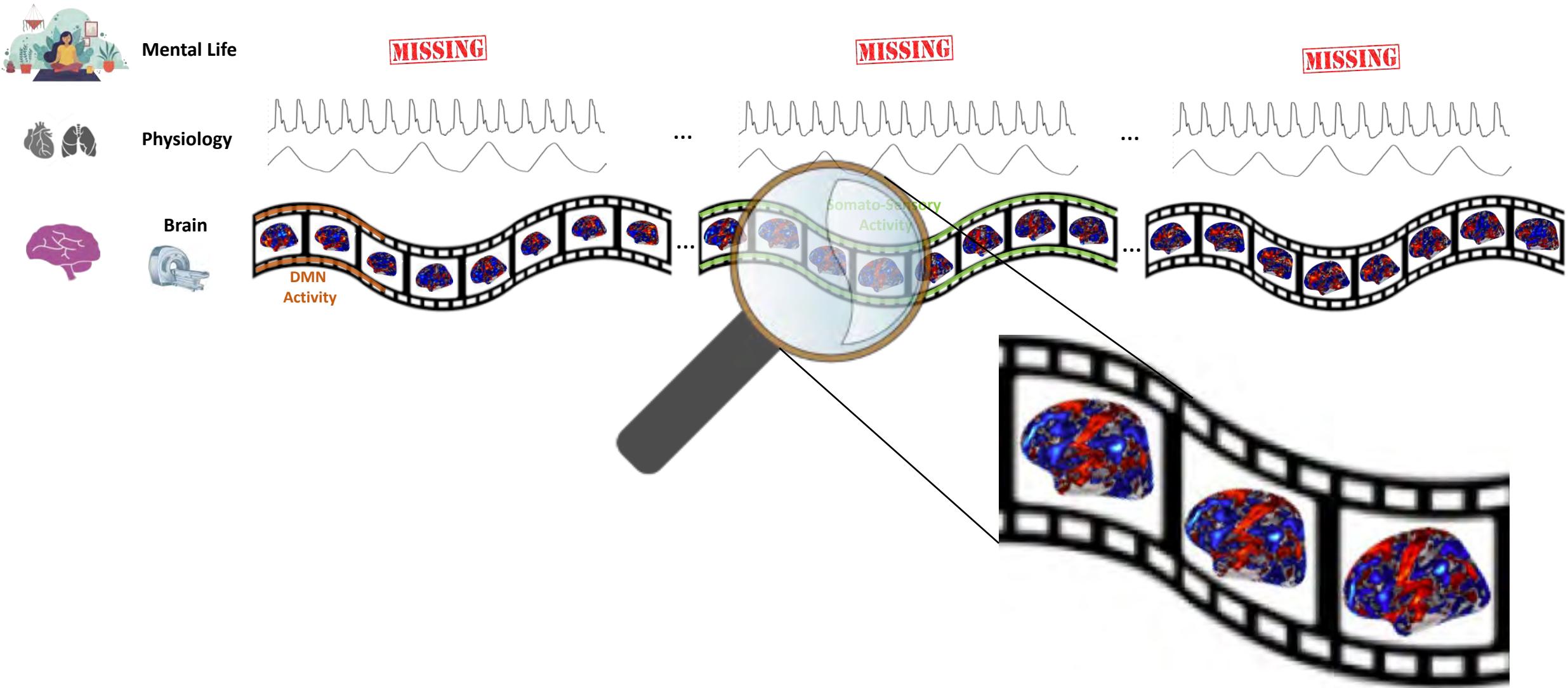


# Introduction: Resting State



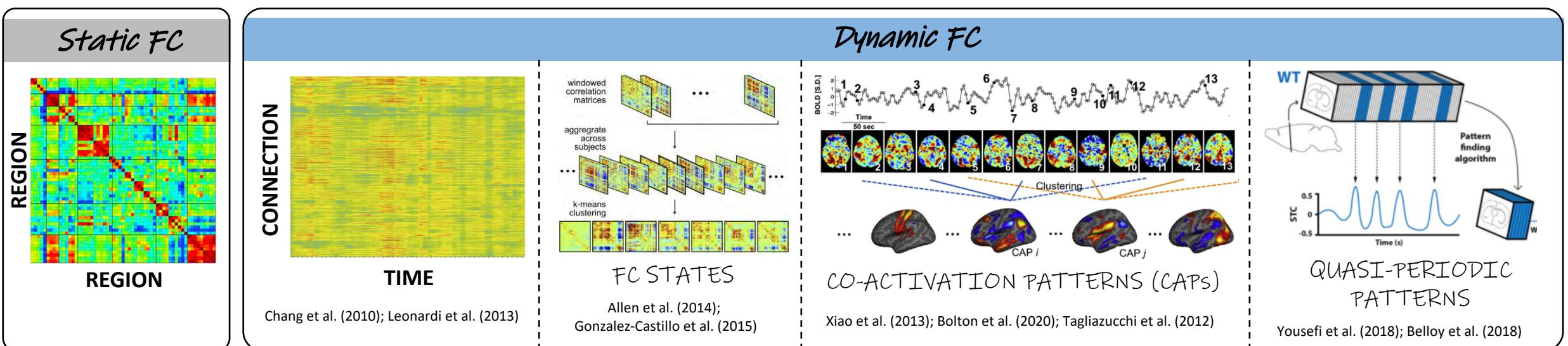
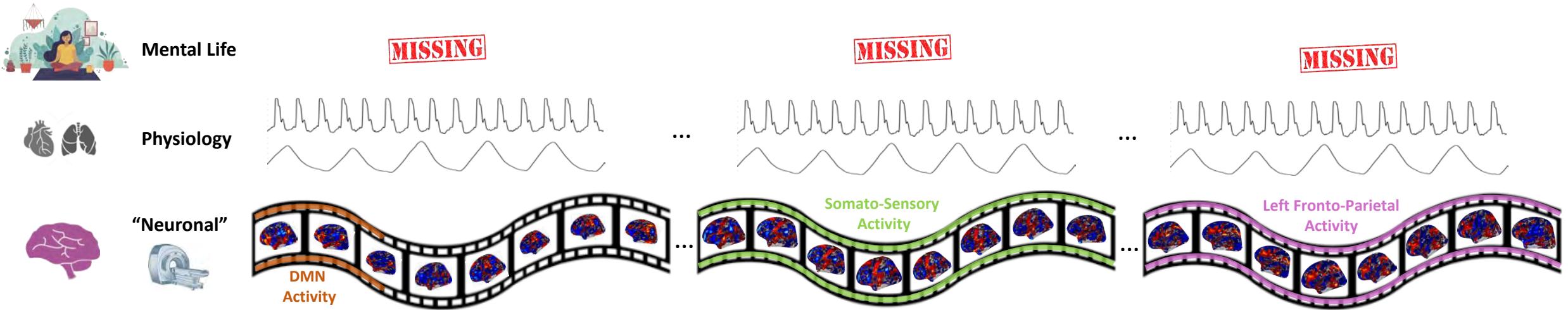


# Introduction: Resting State



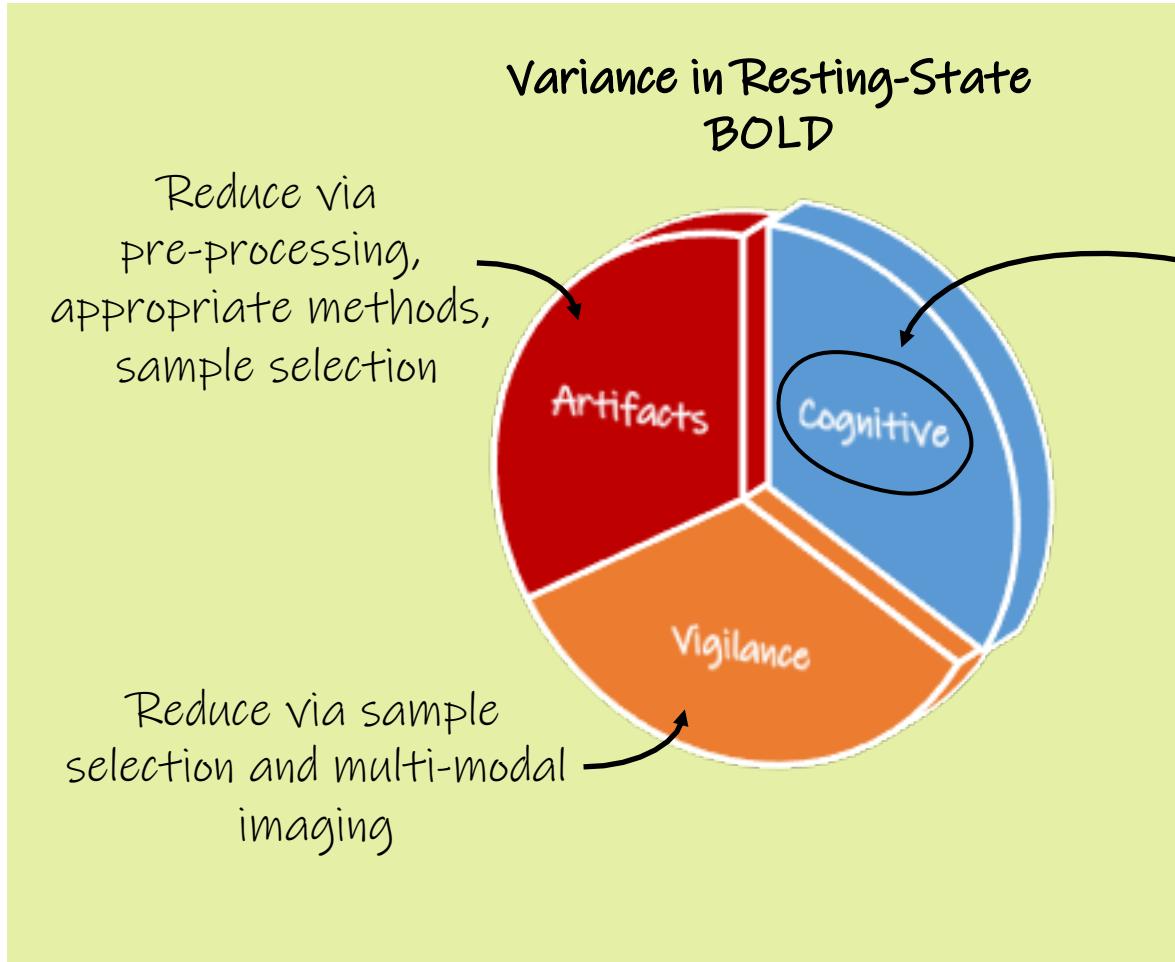


# Introduction: Dynamic FC





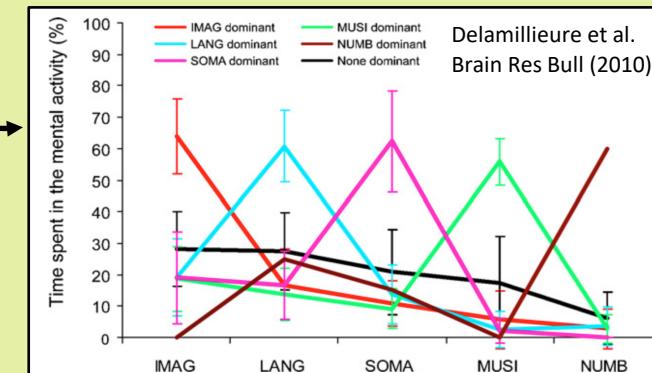
## Introduction - Sources of Variance in Resting-State BOLD



What do subjects do when we ask them to lie still in the scanner and let their mind wander?

- \* Engage in stimuli independent thoughts
- \* Inner speech
- \* Inner seeing
- \* Unsymbolized thinking
- \* Feelings
- \* Have periods of sensory awareness
  - \* Visual   \* Auditory   \* Tactile
- \* General house-keeping functions

There is significant inter-subject variability in the distribution of inner experiences subjects have in the scanner.



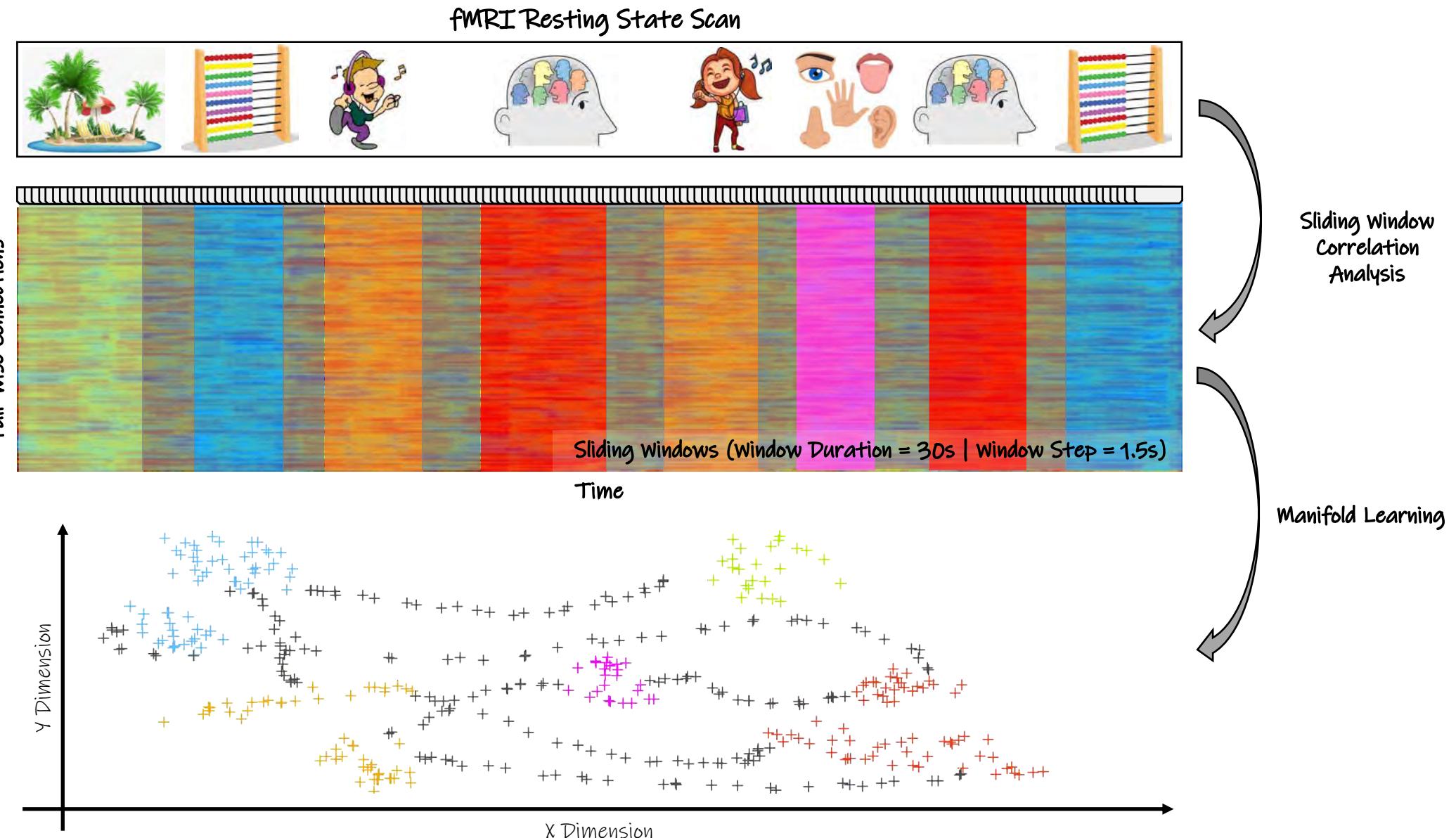


# Linking dynamic FC to ongoing cognition – Methods Development: Framework

Mental Life Perspective

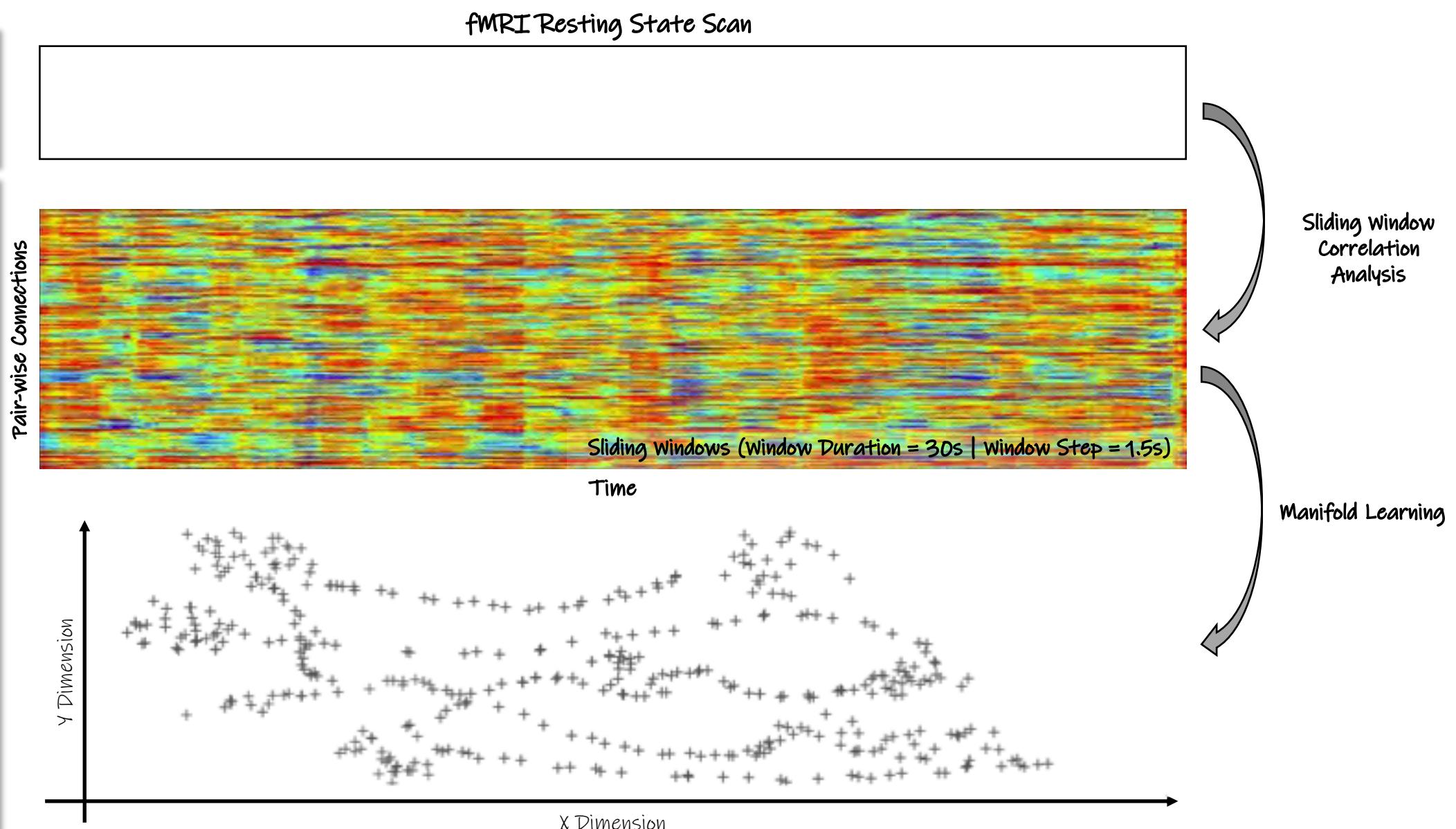
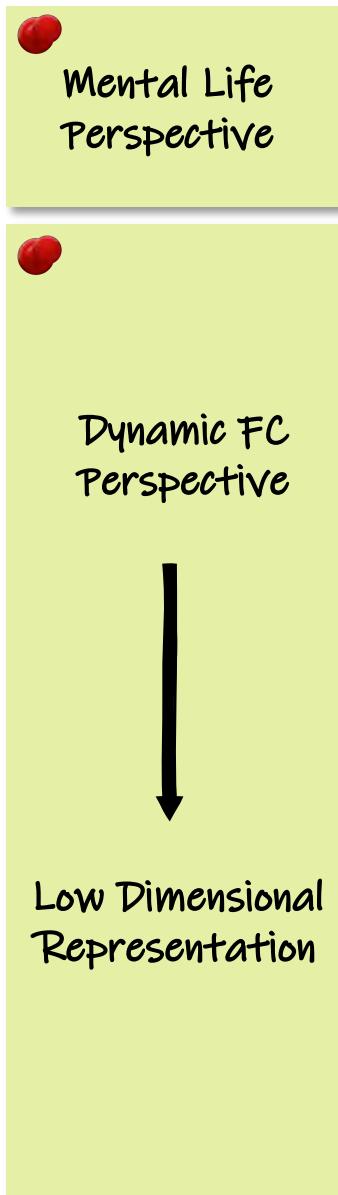
Dynamic FC Perspective

Low Dimensional Representation





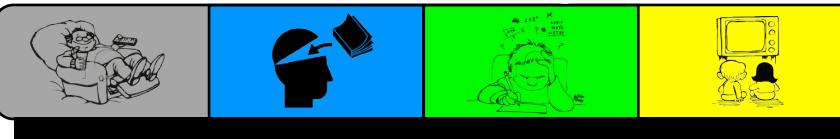
# Linking dynamic FC to ongoing cognition – Methods Development: Framework





# Linking dynamic FC to ongoing cognition - Methods Development: Datasets

## Multi-Task Dataset | Testing



(180")

**Instructions (12")**

REST

MEMORY

VIDEO

MATH

MEMORY

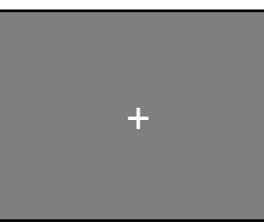
**Total Run Duration 25'24"**

REST

MATH

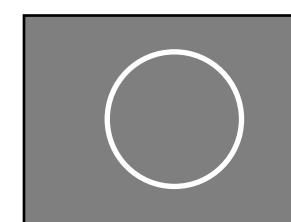
VIDEO

REST



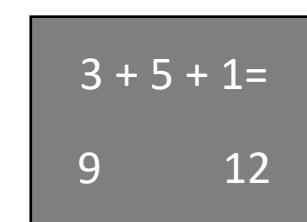
Passively fixate on  
the crosshair on the  
center of the screen

MEMORY



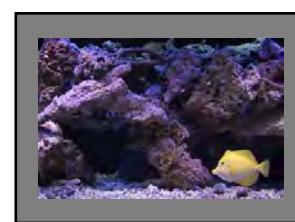
Press button when  
shape in screen is  
same as two before.

MATH



Select the correct  
answer from the two  
available options.

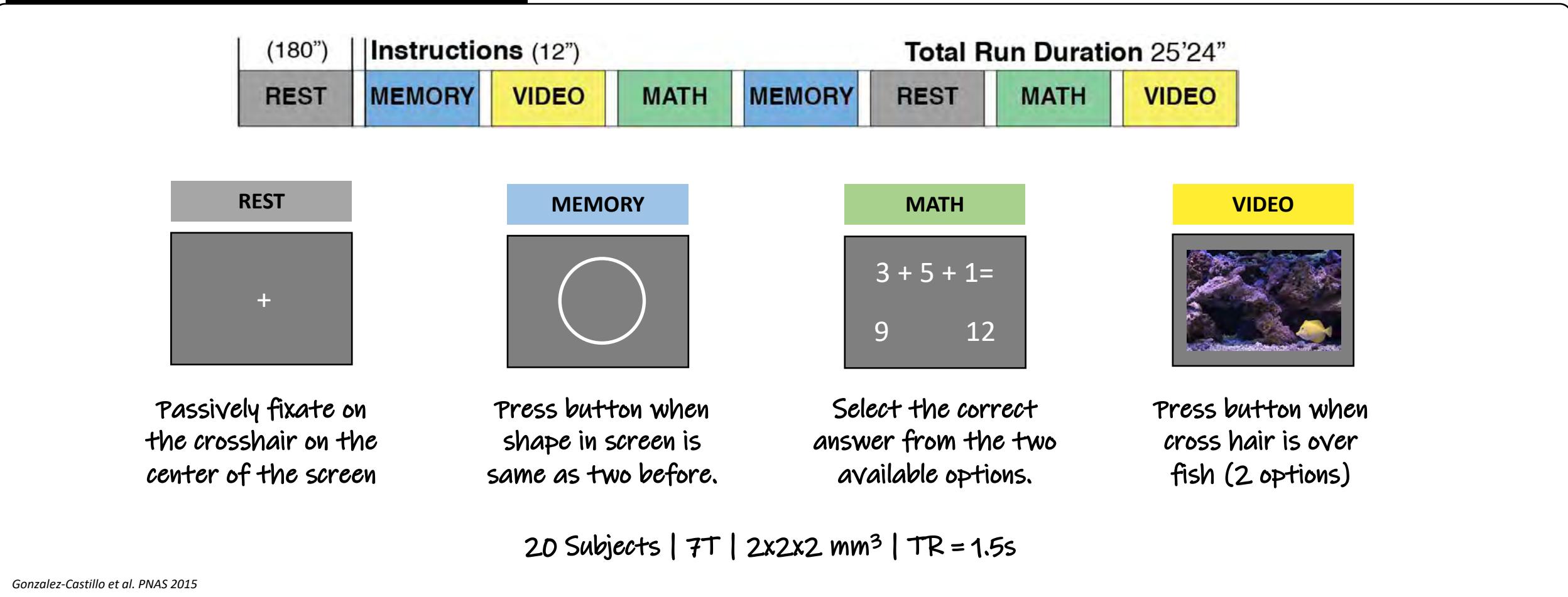
VIDEO



Press button when  
cross hair is over  
fish (2 options)

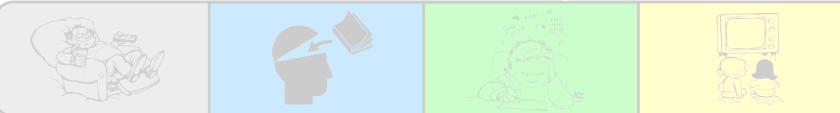
20 Subjects | 7T | 2x2x2 mm<sup>3</sup> | TR = 1.5s

## Continuous Rest Dataset | Application





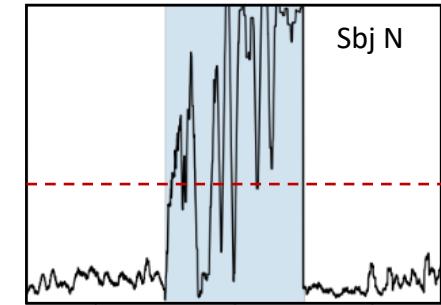
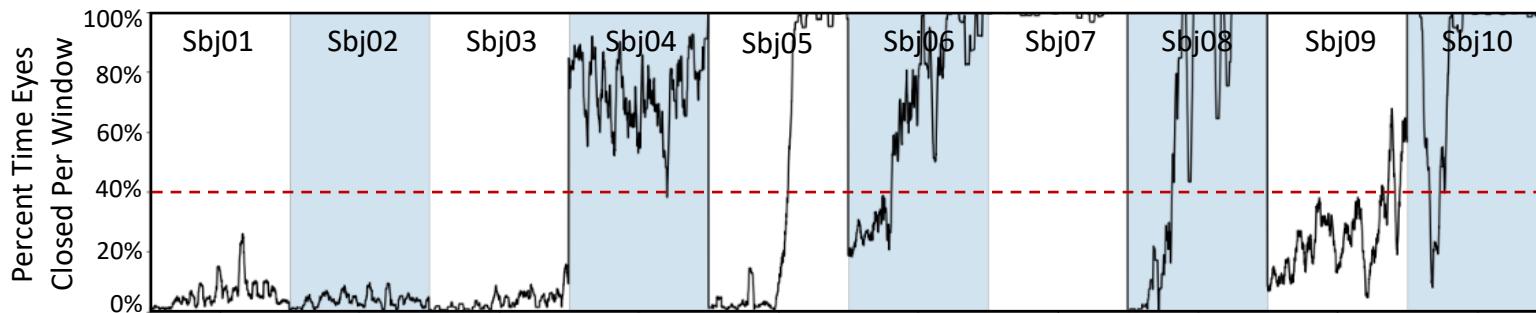
## Multi-Task Dataset | Testing



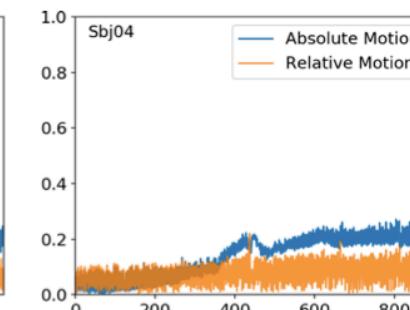
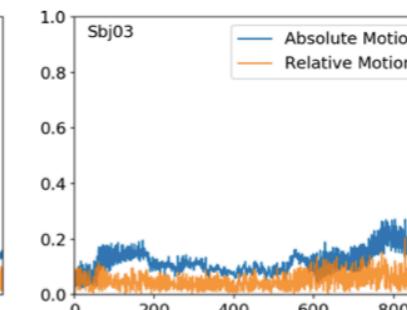
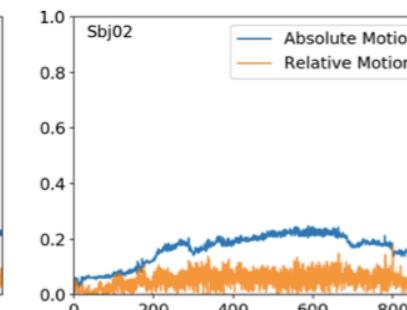
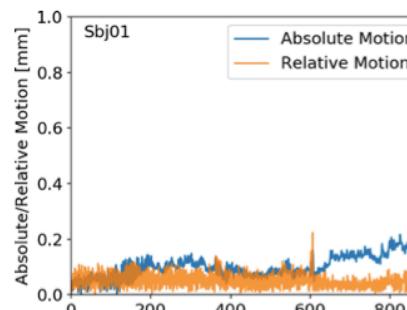
## Continuous Rest Dataset | Application



- 175 Subjects, each with four 15-min long, eyes-open resting state scans with concurrent eye tracking recordings.
- Subjects that stayed awake during the whole rest scan (based on eye tracking traces).



- Of these, we focused on the 20 subjects with the least amount of motion.



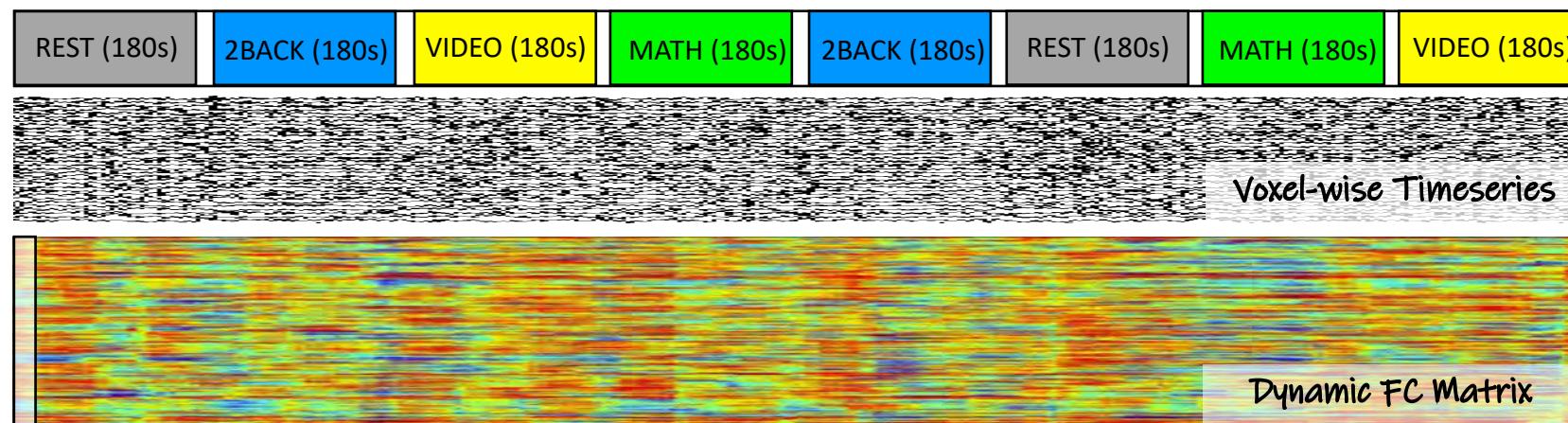
Average absolute motion was  $0.27 \pm 0.18$  mm, and average relative volume-to-volume motion was  $0.10 \pm 0.07$  mm.



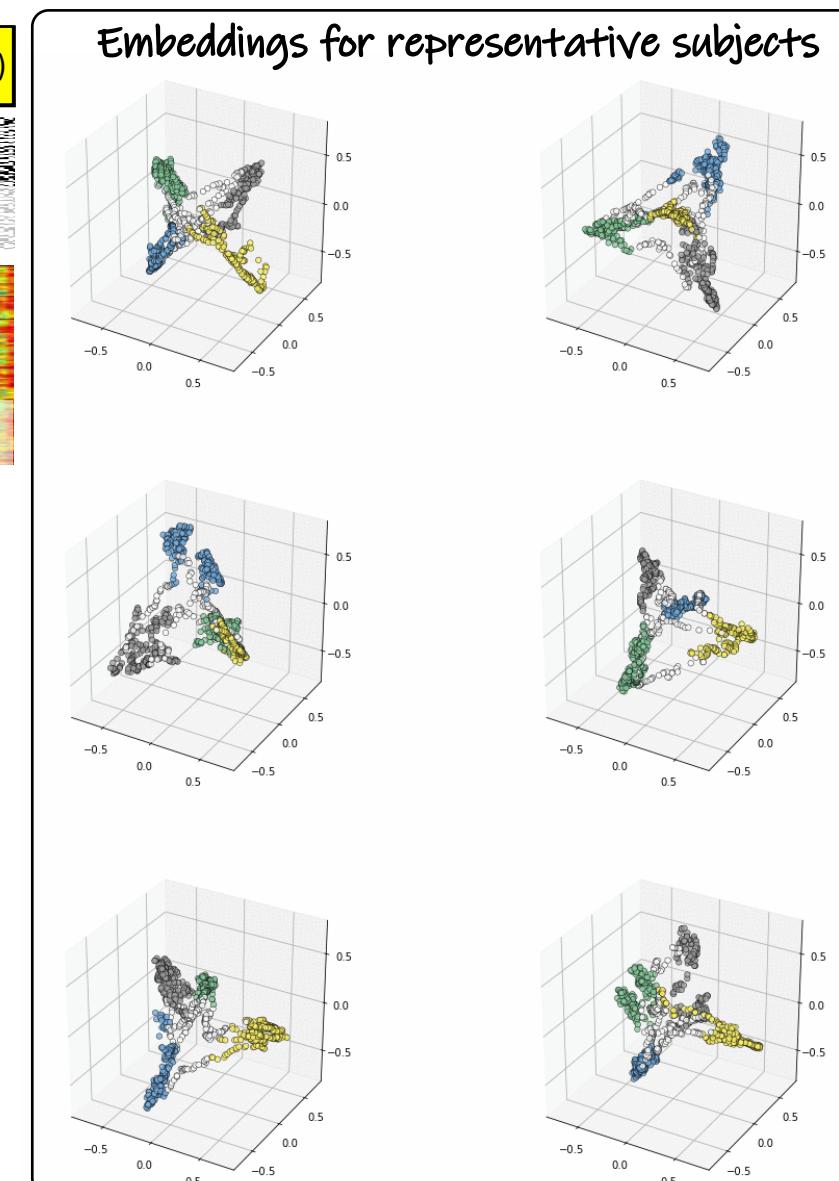
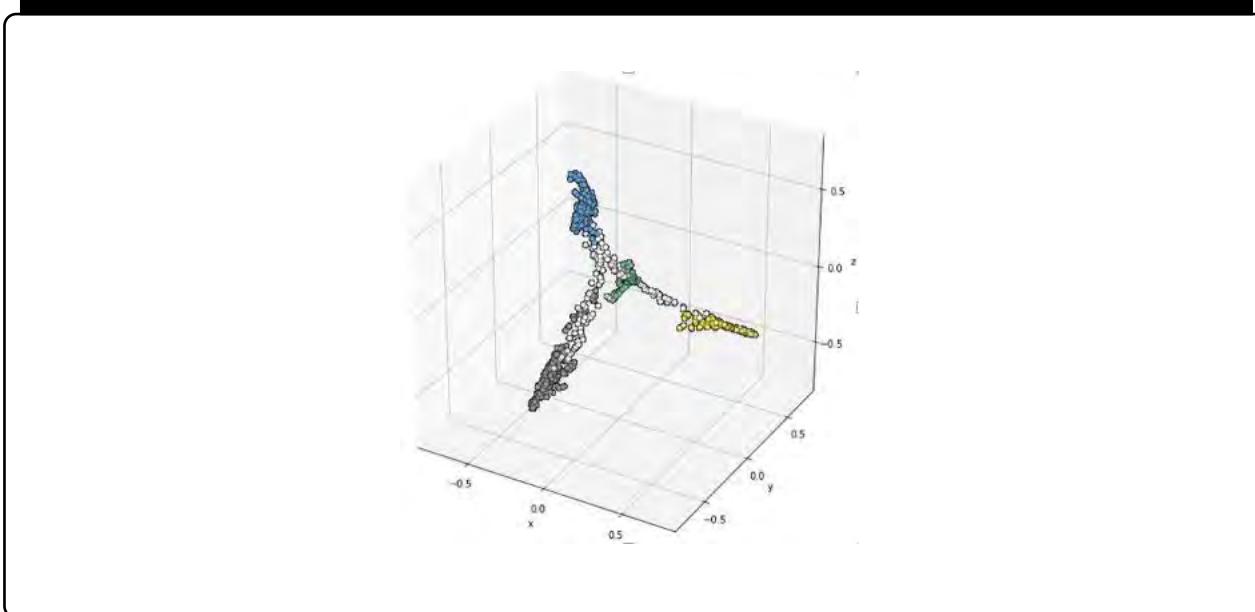
# Linking dynamic FC to ongoing cognition – Methods Evaluation on Multi-Task Dataset

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## 1. Generate Low Dimensional Representations of dFC Matrix

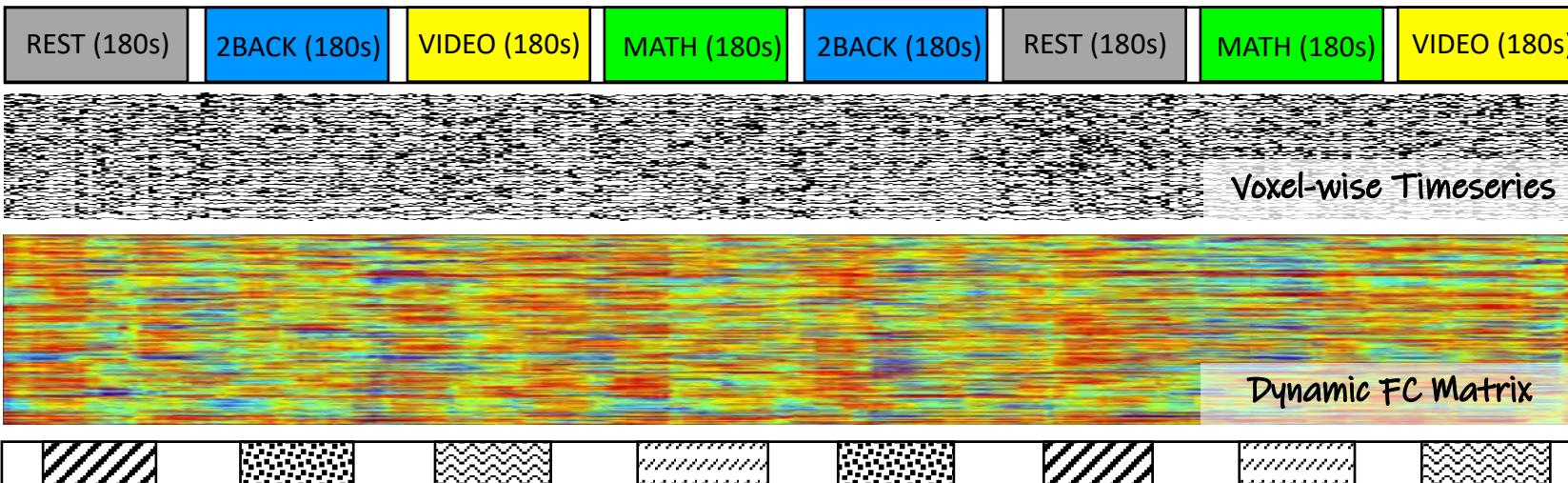




# Linking dynamic FC to ongoing cognition – Methods Evaluation on Multi-Task Dataset

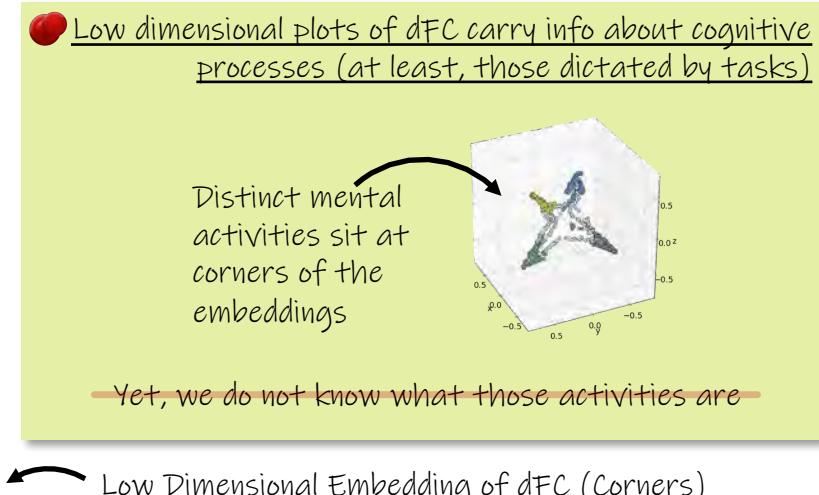
Section on Functional Imaging Methods

NIH National Institute of Mental Health

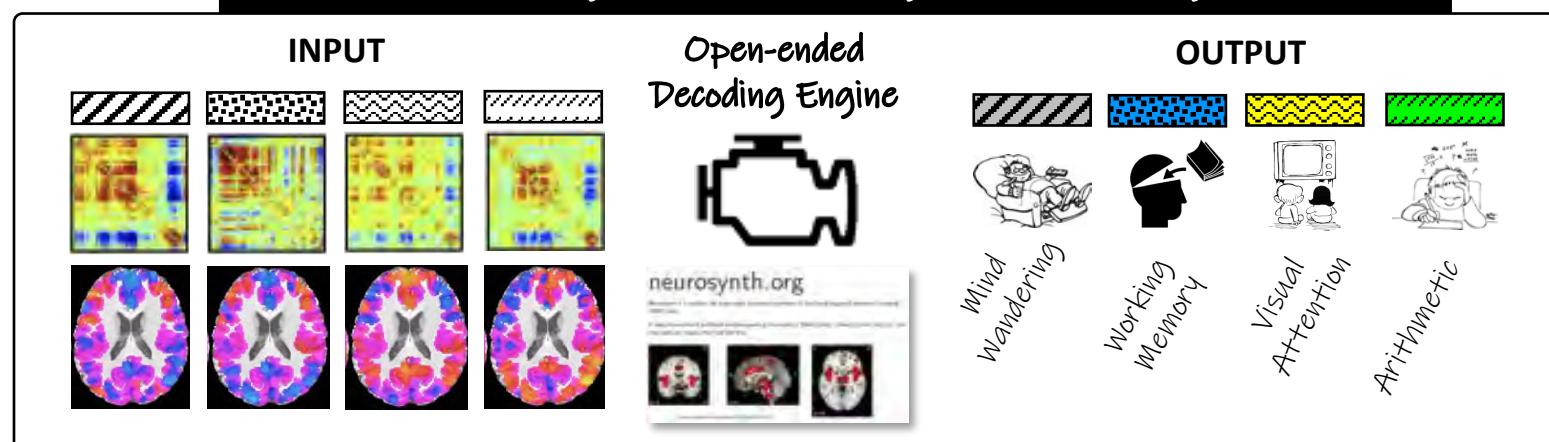


Voxel-wise Timeseries

Dynamic FC Matrix



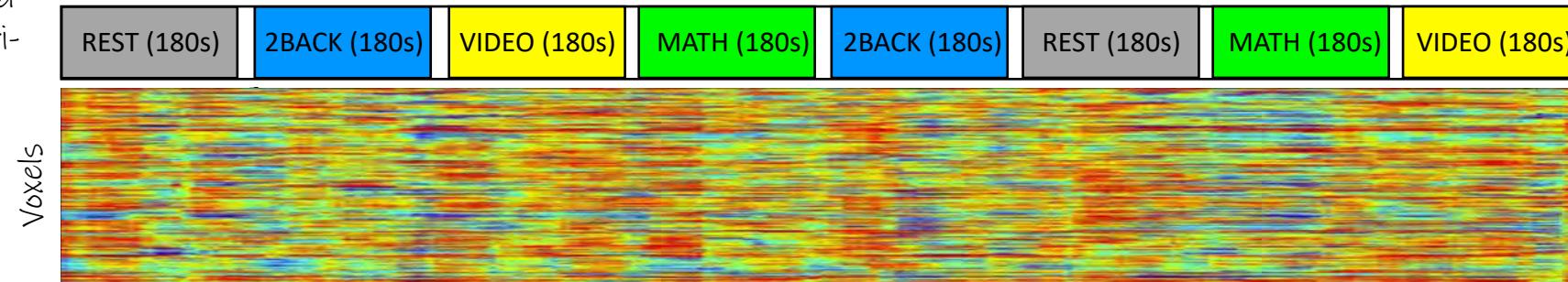
## 2. Decode Cognitive Processes aligned with each segment





# Linking dynamic FC to ongoing cognition – Methods Evaluation on Multi-Task Dataset

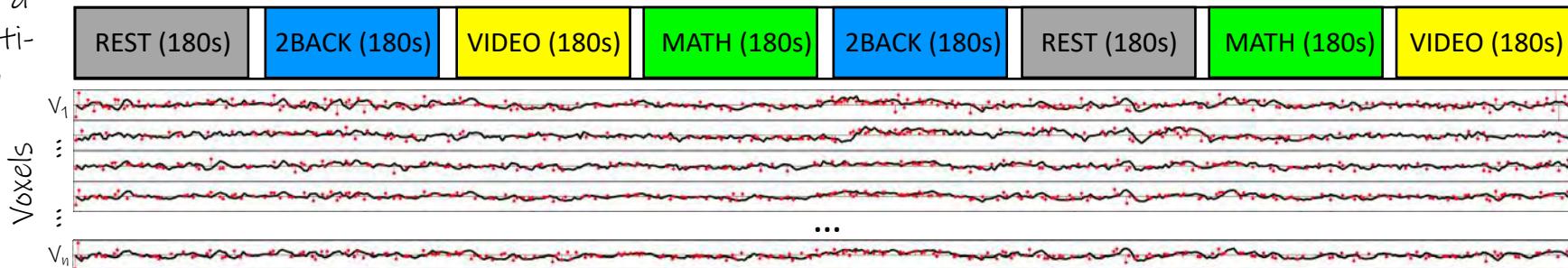
1. We switch from a dFC-view to a multi-variate activity view.





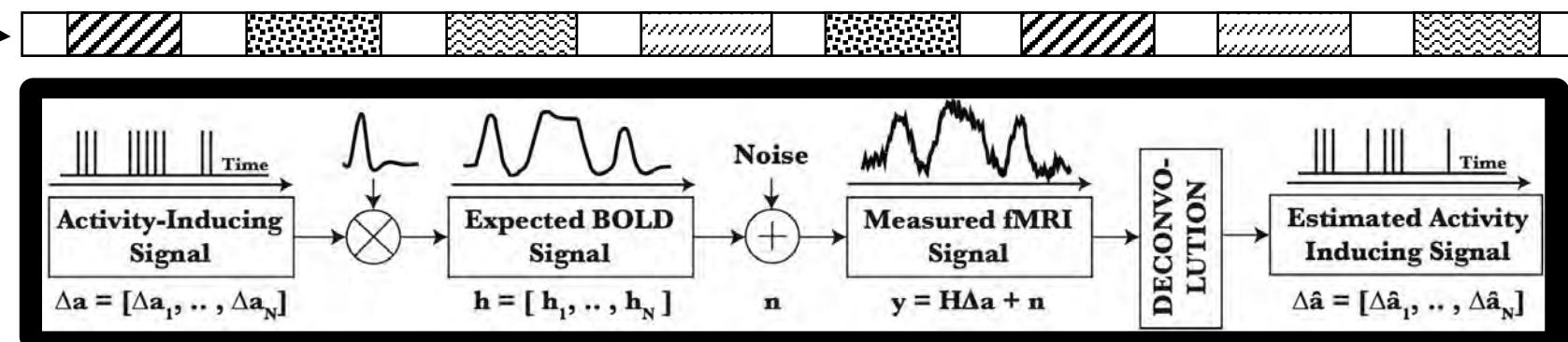
# Linking dynamic FC to ongoing cognition – Methods Evaluation on Multi-Task Dataset

1. We switch from a dFC-view to a multi-variate activity view.



2. We apply deconvolution to find most prominent HRF-inducing events

3. We rely on the dFC-based scan partitioning.

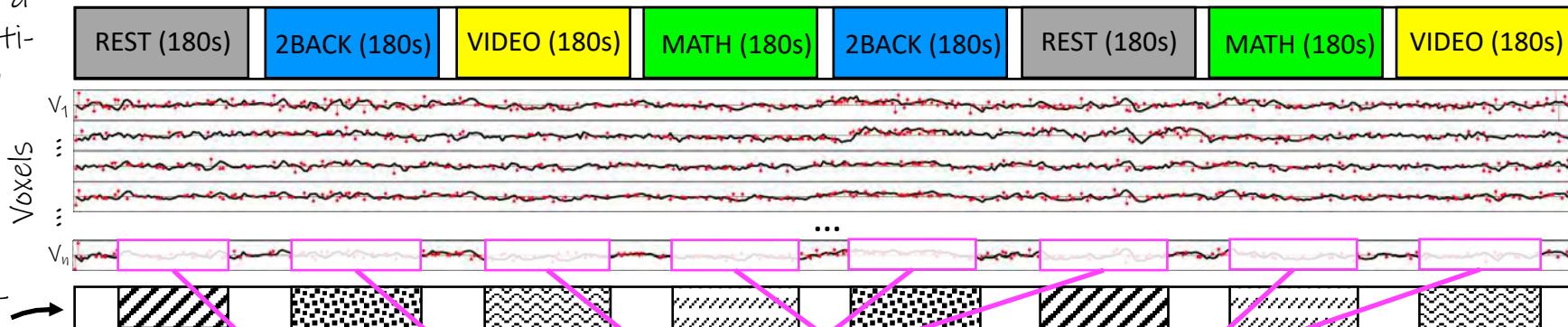


Hemodynamic Deconvolution – Find Most Prominent Activity Inducing Events  
(SPFM; Caballero-Gaudes et al. HMB 2011)

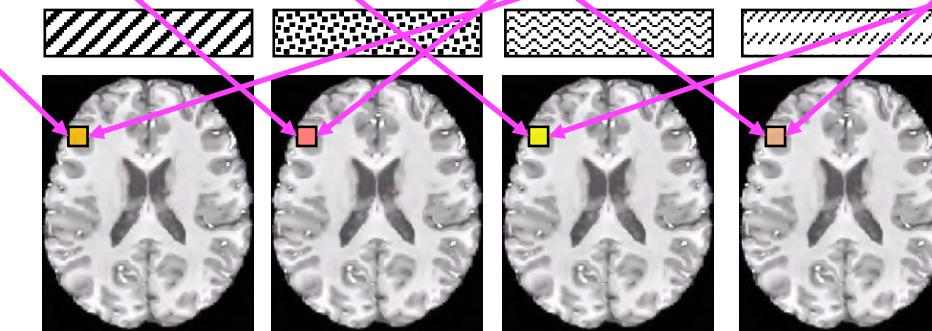


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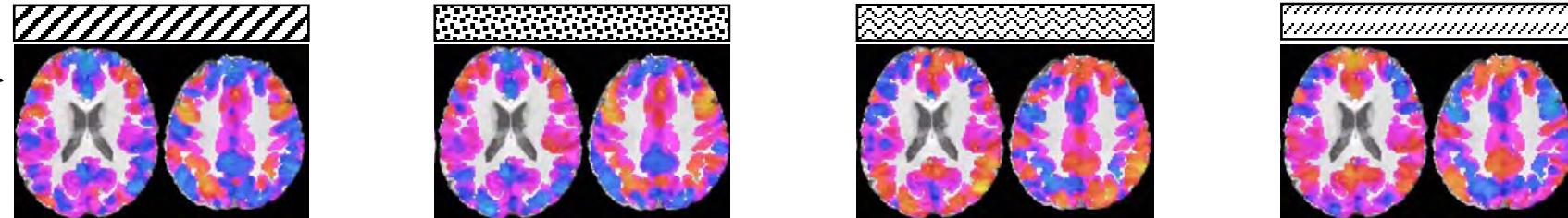
1. We switch from a dFC-view to a multi-variate activity view.



3. We rely on the dFC-based scan partitioning.



4. Create "activity" maps per segment via deconvolution

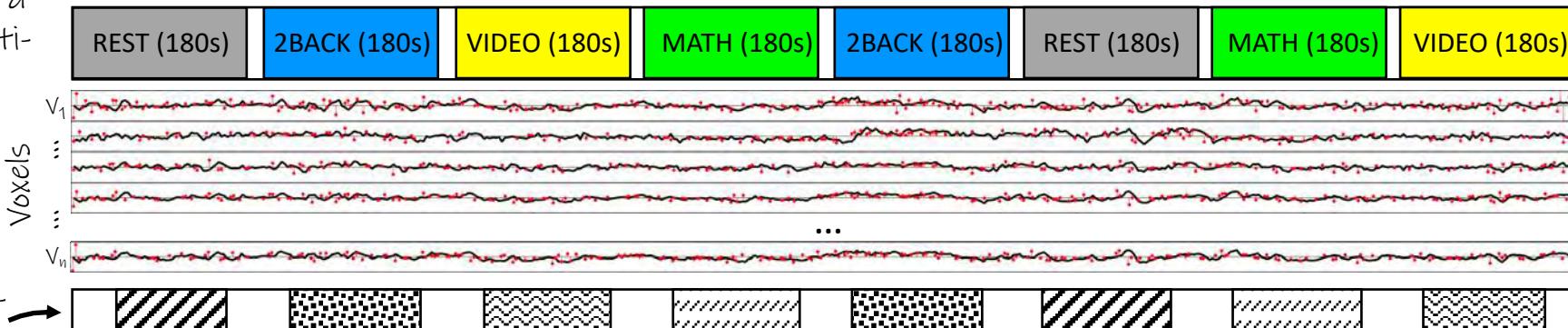


2. We apply deconvolution to find most prominent HRF-inducing events



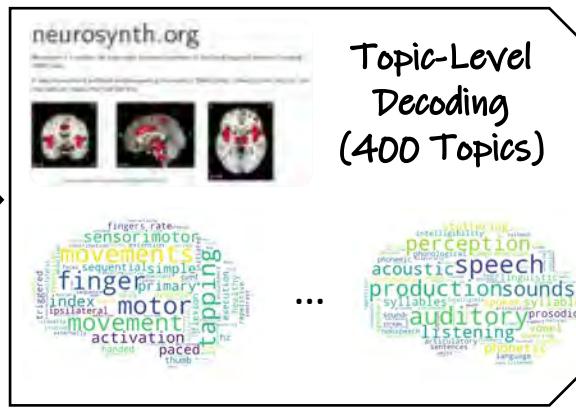
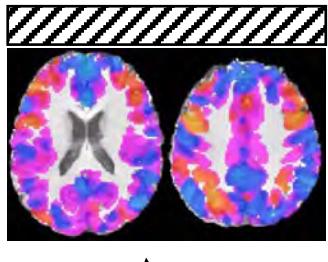
# Linking dynamic FC to ongoing cognition – Methods Evaluation on Multi-Task Dataset

**1.** We switch from a dFC-view to a multi-variate activity view.



**2.** We apply deconvolution to find most prominent HRF-inducing events

**3.** We rely on the dFC-based scan partitioning.



001	$R=0.23$	numbers, number, size, numerical, arithmetic, processing, calculation, magnitude, distance, ...
002	$R=0.22$	perception, acoustic, speech, production, sounds, auditory, listening, ...
003	$R=0.18$	emotion, negative, neutral, processing, emotional, positive, regulation, ...
...	...	...
400	$R=-0.3$	motor, primary, imagery, control, movement, sensorimotor, movement, sensorimotor, movement, ...

**4.** Create "activity" maps per segment via deconvolution

**5.** Use Neurosynth's reverse-inference abilities to obtain cognitive topics that most highly relate to each "activity" map

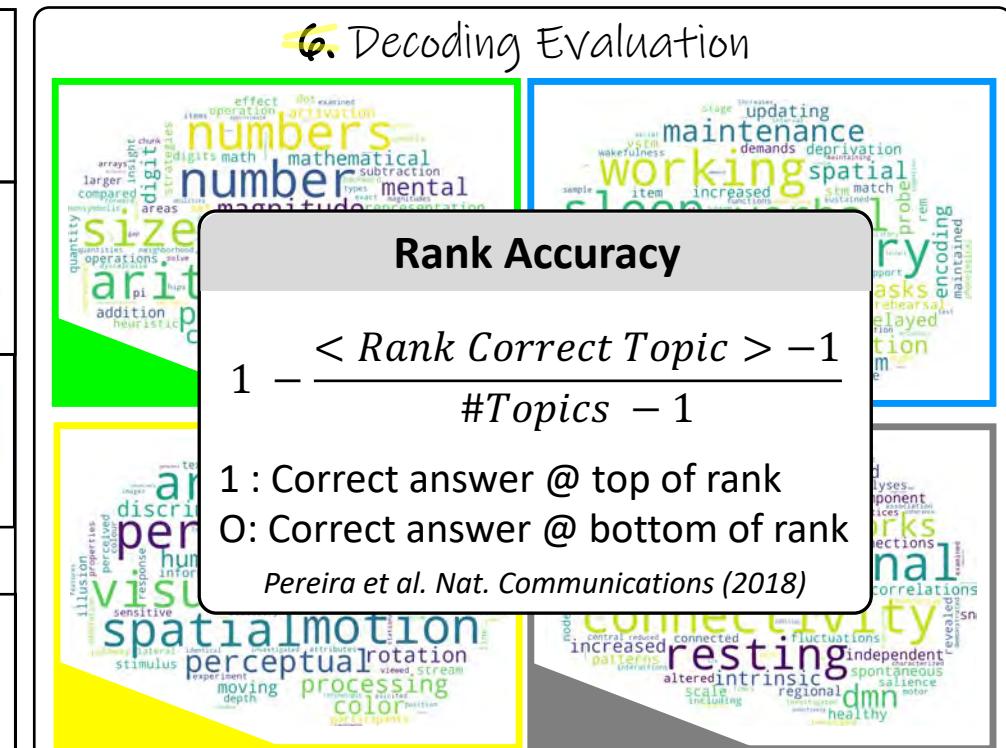
**6. Decoding Evaluation**

**Rank Accuracy**

$$1 - \frac{\langle \text{Rank Correct Topic} \rangle - 1}{\# \text{Topics} - 1}$$

- 1 : Correct answer @ top of rank  
0: Correct answer @ bottom of rank

Pereira et al. Nat. Communications (2018)



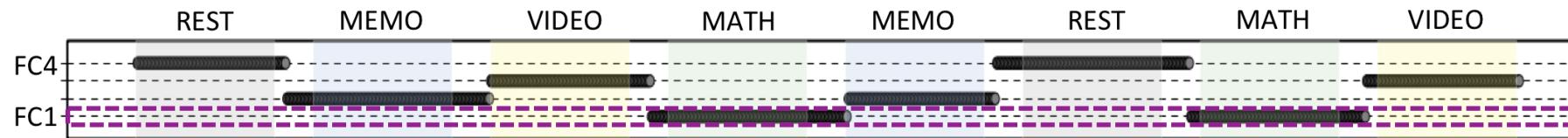


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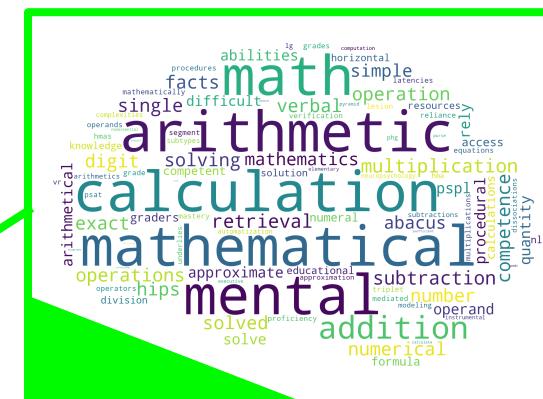
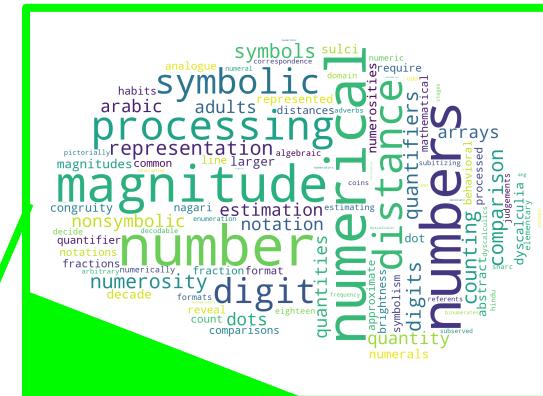
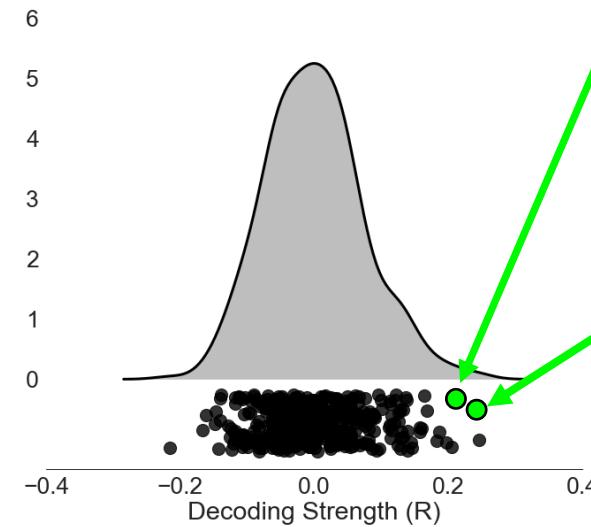
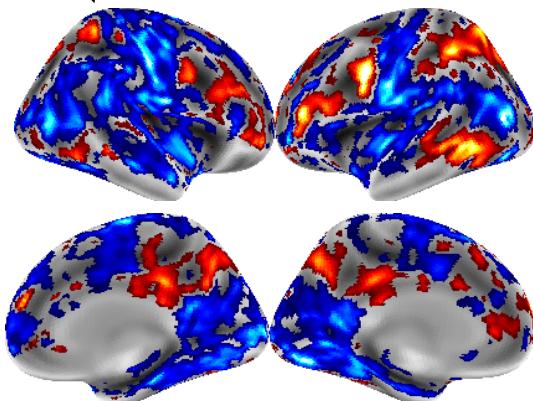
Section on Functional Imaging Methods FIM

National Institute of Mental Health

Automatic Scan Segmentation  
based on dynamic-FC



"Activity" Map  
for FC1



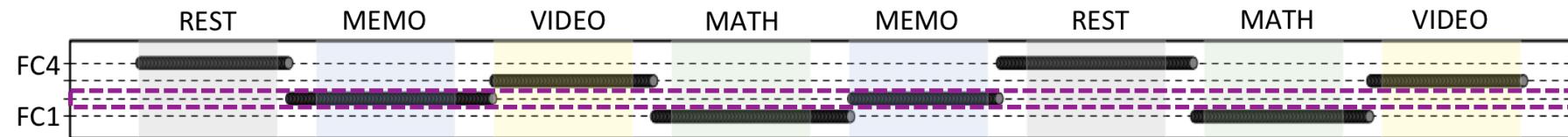


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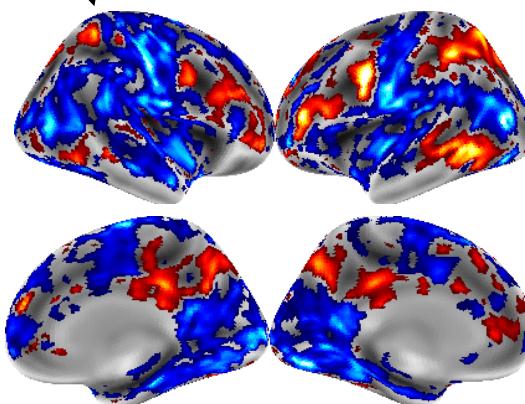
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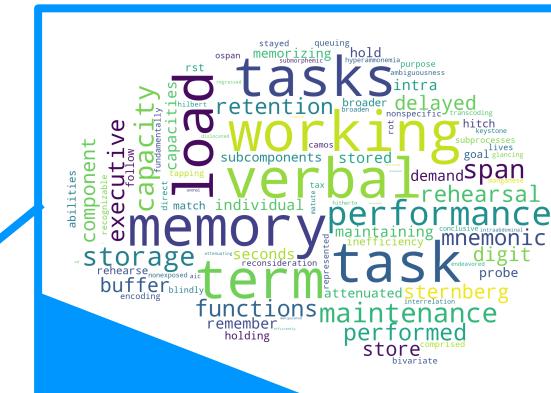
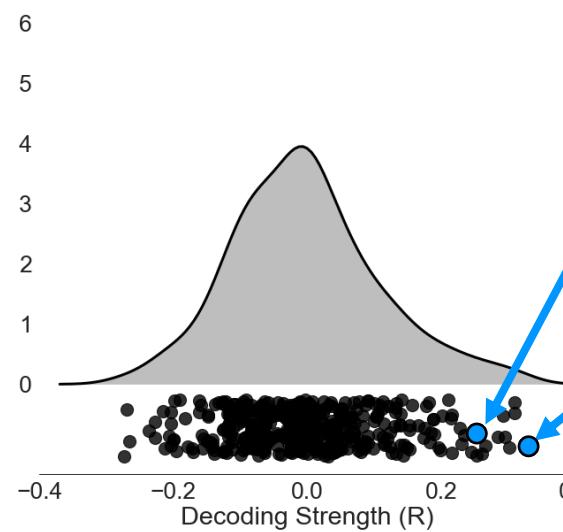
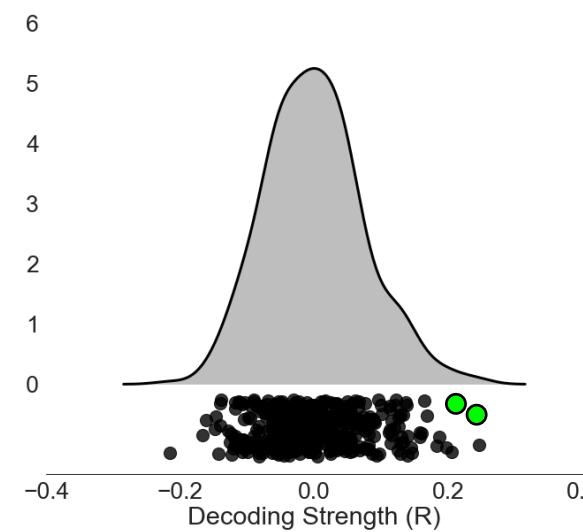
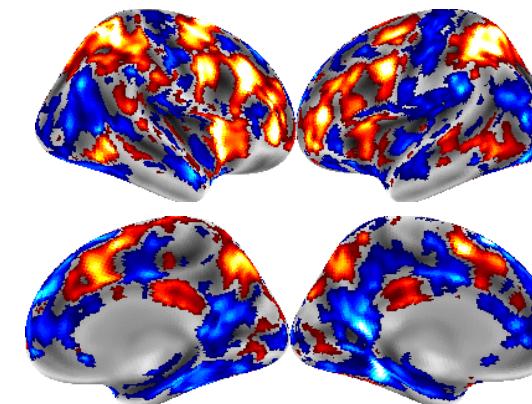
Automatic Scan Segmentation  
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"Activity" Map  
for FC1



FC2 - "Activity" Map





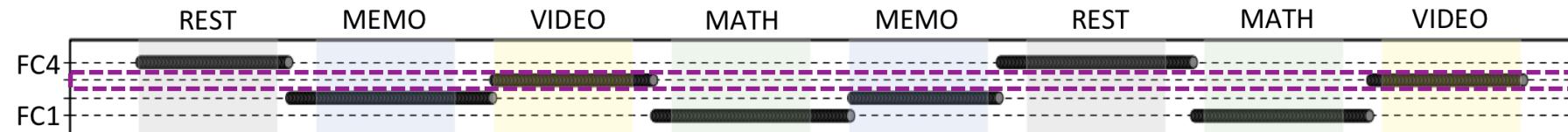
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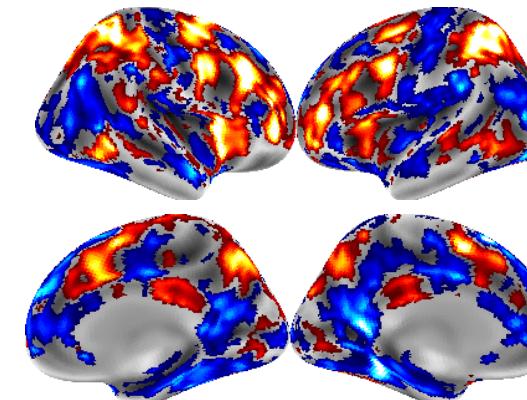
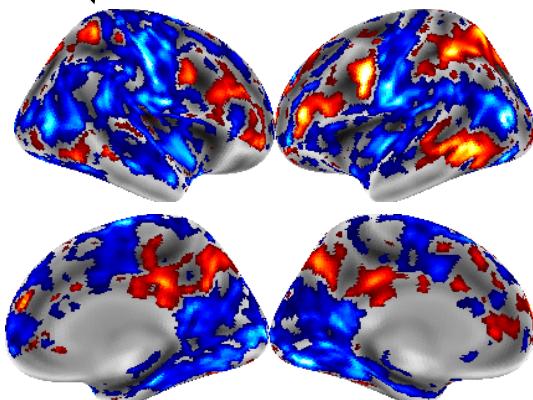
National Institute  
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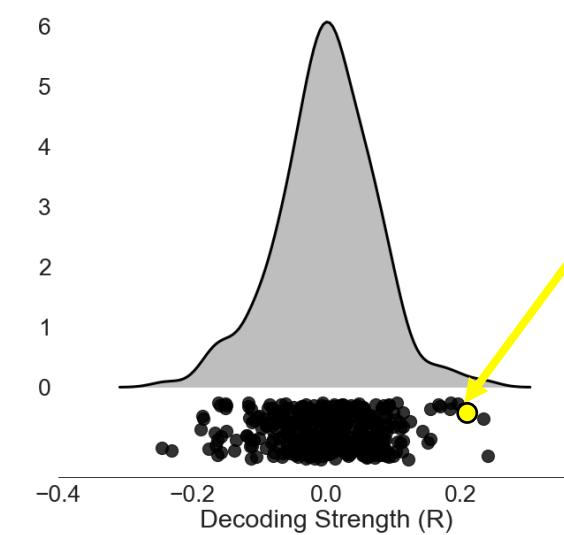
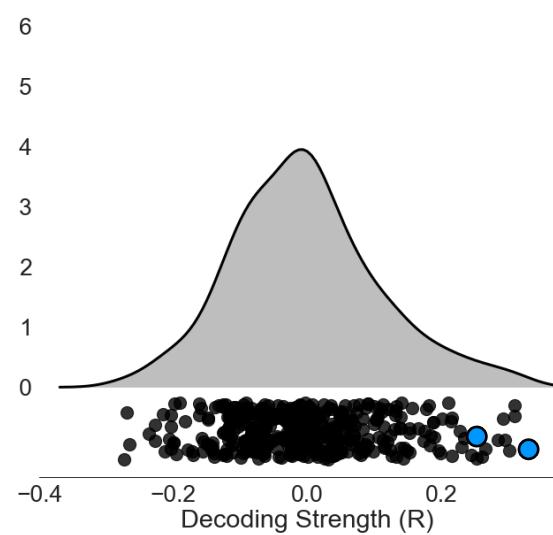
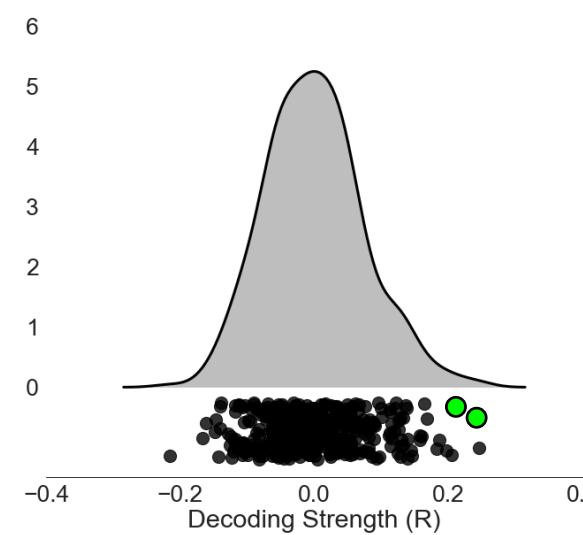
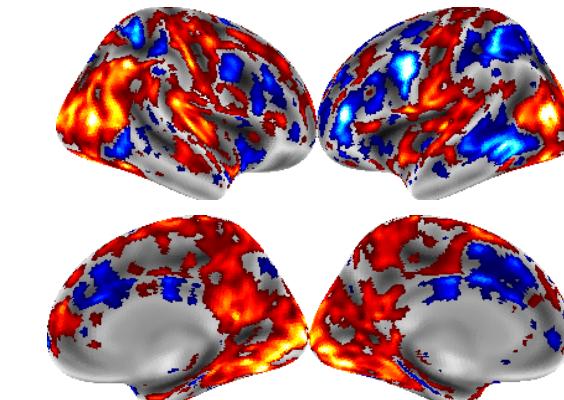
"Activity" Map  
for FC1



FC2 - "Activity" Map



FC3 - "Activity" Map



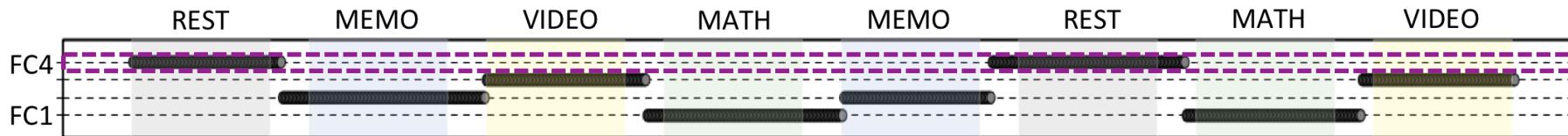


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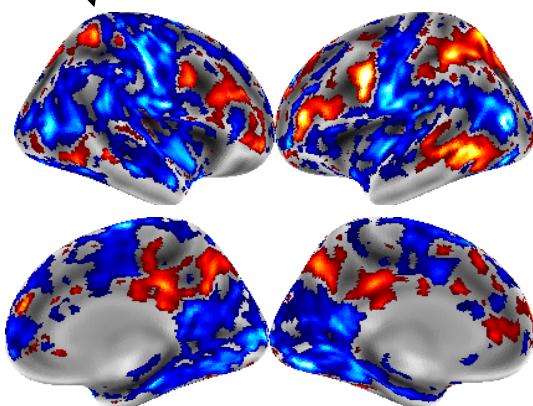
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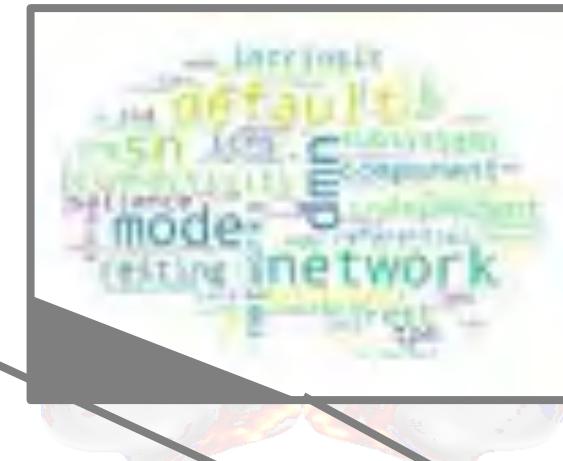
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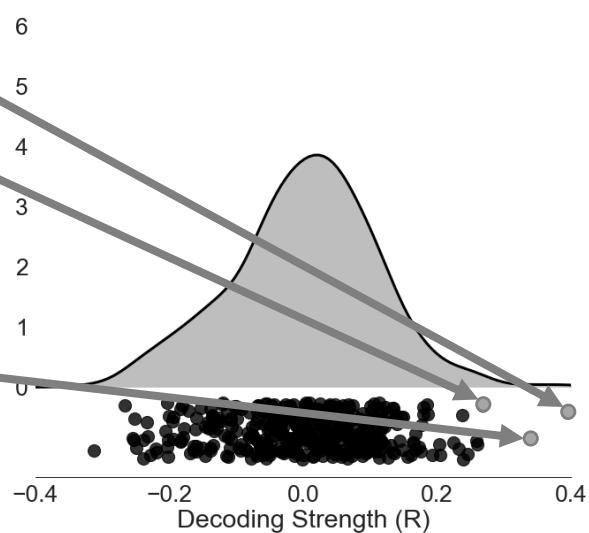
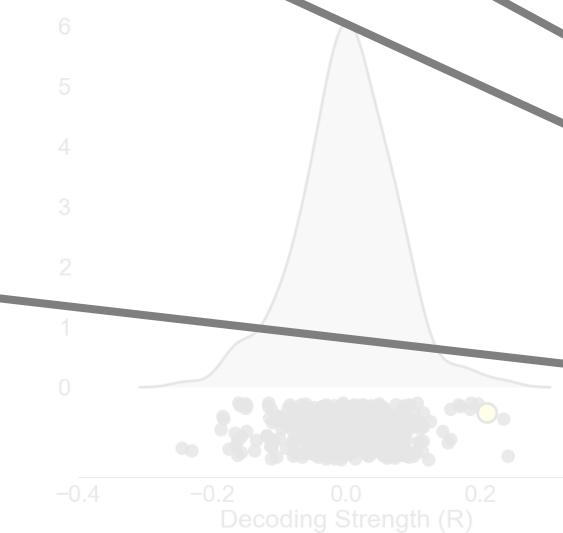
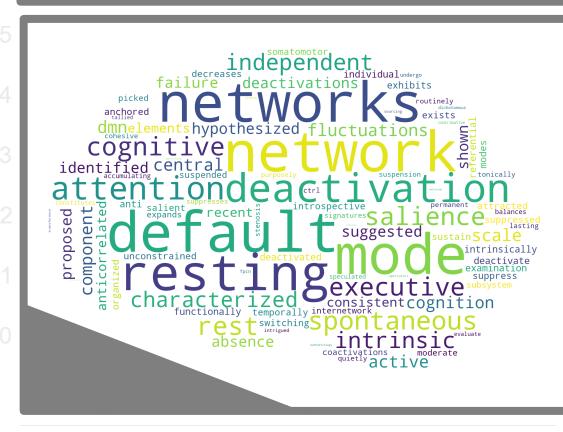
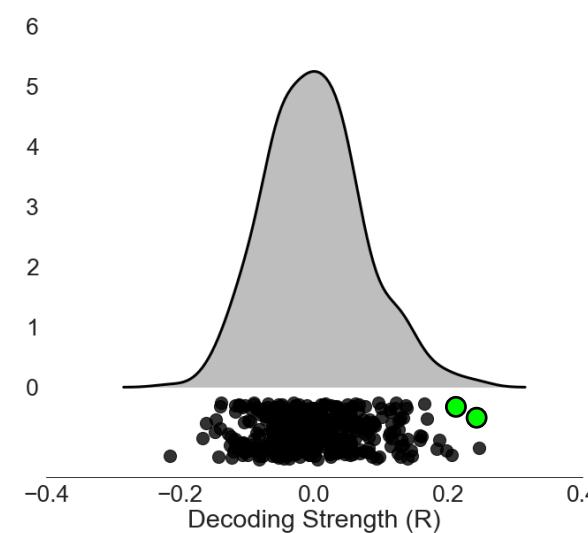
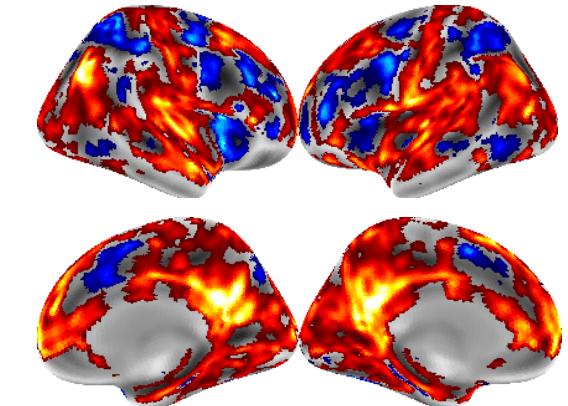
"Activity" Map  
for FC1



FC2 - "Activity" Map

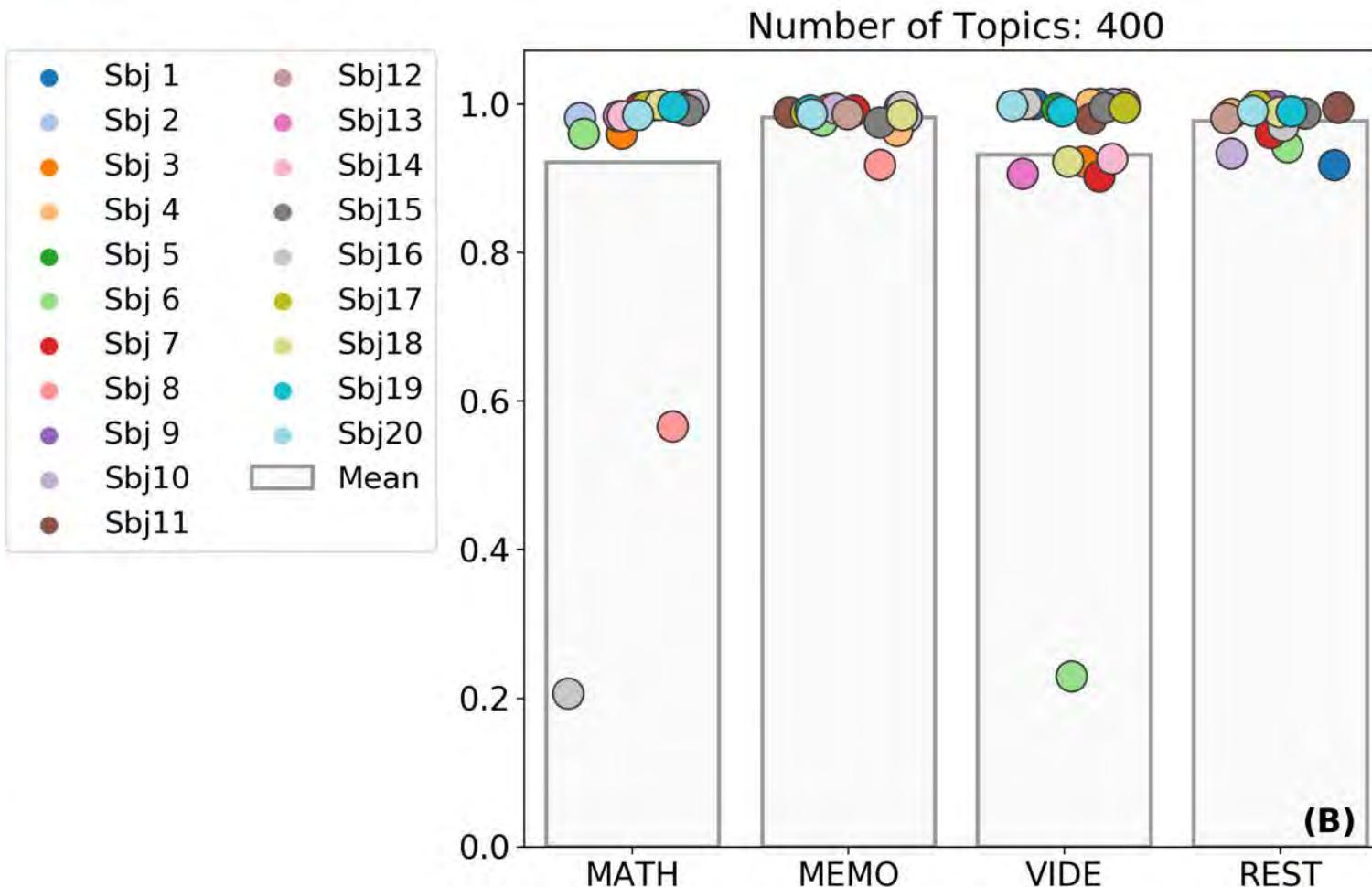


FC4 - "Activity" Map





## Group Level Decoding Accuracy Results

**Rank Accuracy**

$$1 - \frac{\langle \text{Rank Correct Topic} \rangle - 1}{\# \text{Topics} - 1}$$

1 : Correct answer @ top of rank

0: Correct answer @ bottom of rank

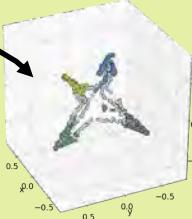
Pereira et al. Nat. Communications (2018)



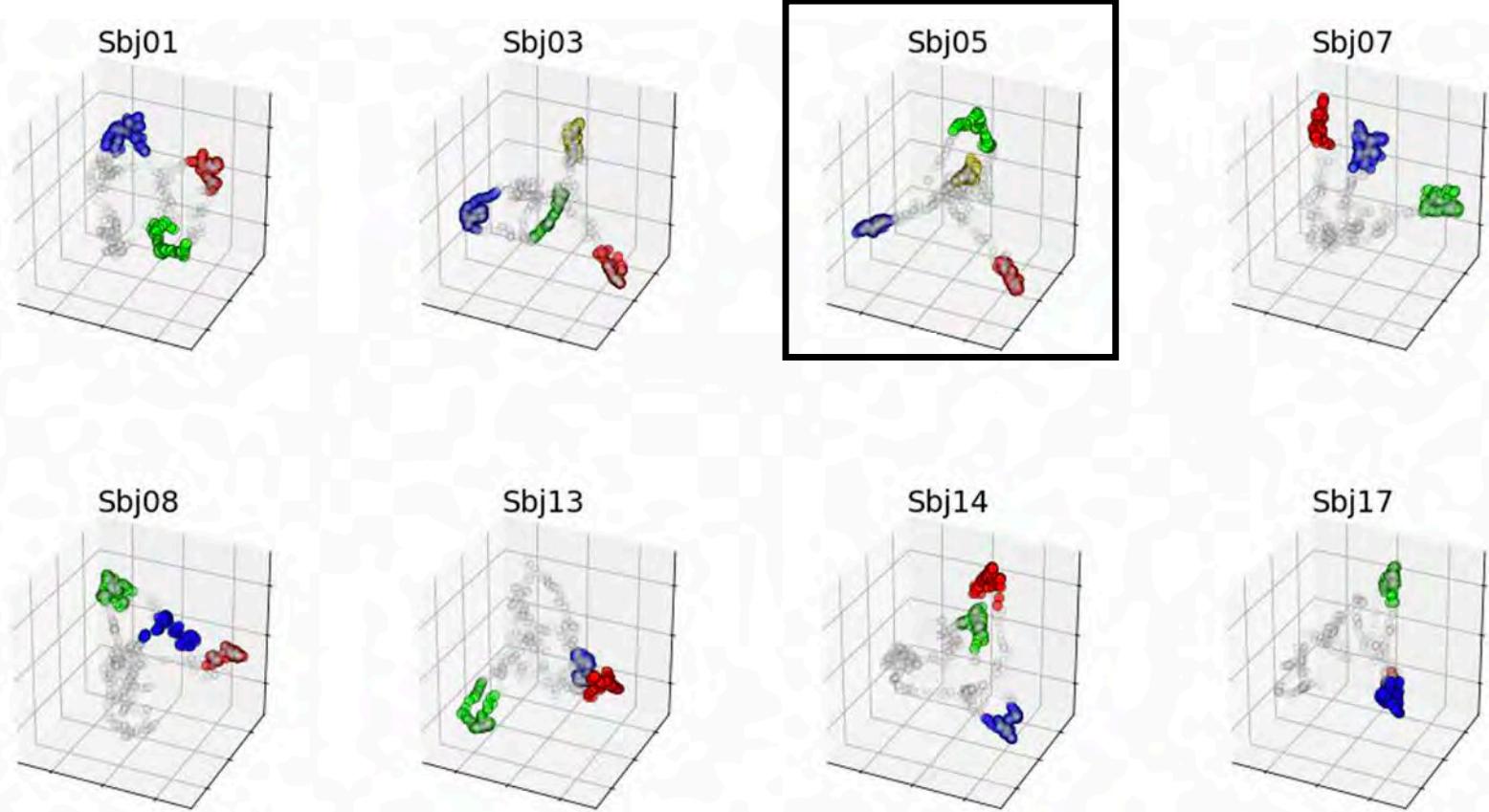
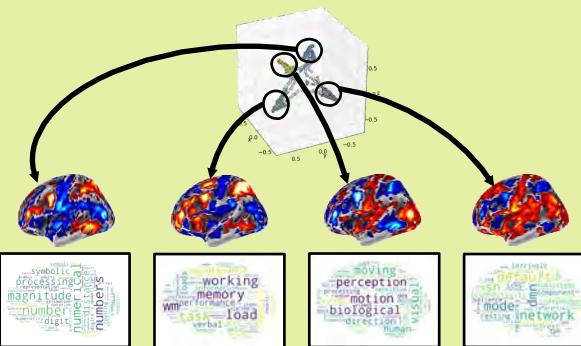
# Linking dynamic FC to ongoing cognition - Application to Resting-State Dataset

Low dimensional plots of dFC carry info about cognitive processes (at least, those dictated by tasks)

Distinct mental activities sit at corners of the embeddings

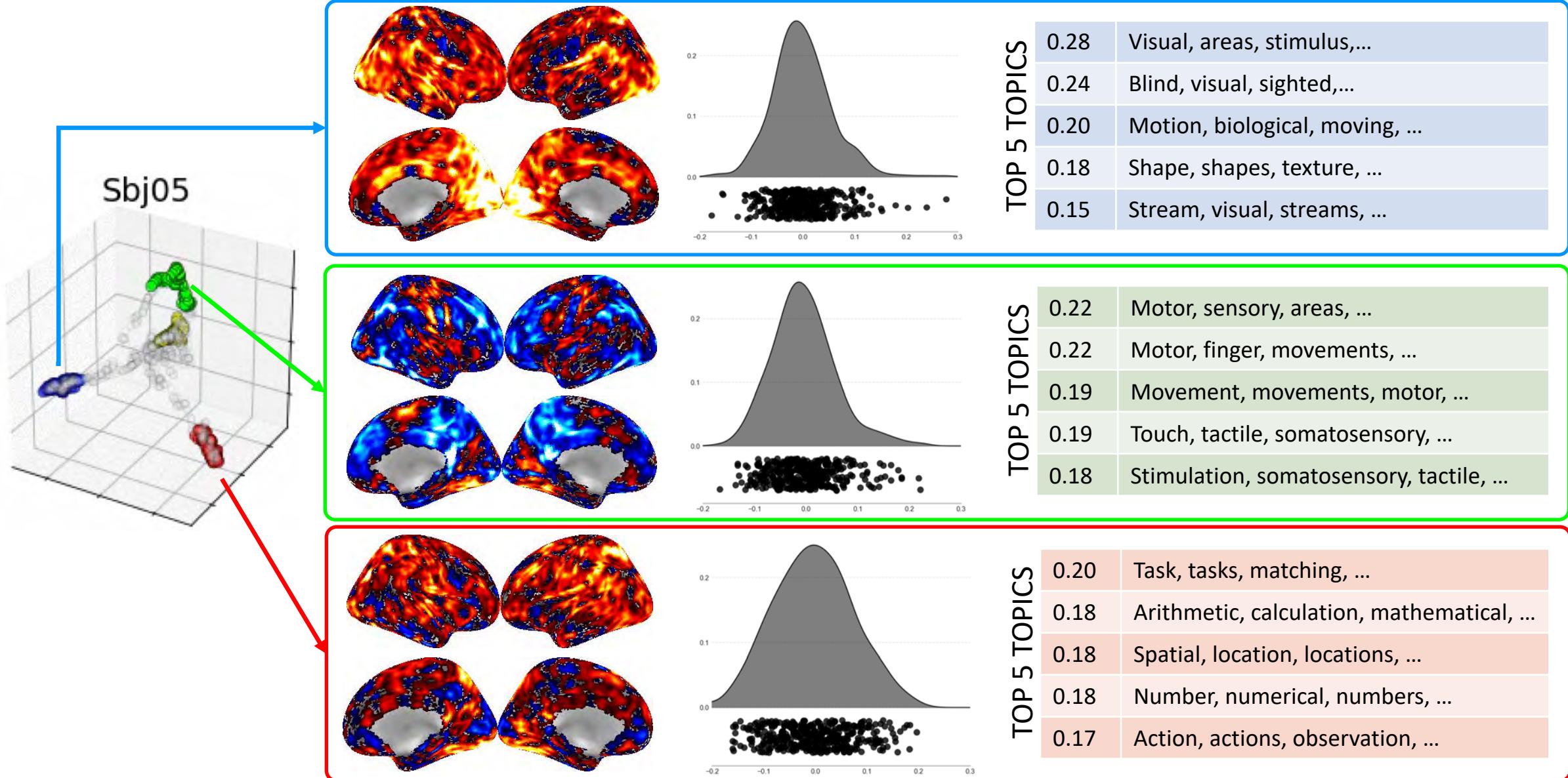


Deconvolution + Neurosynth to infer the nature of the main cognitive processes ongoing during each segment





# Linking dynamic FC to ongoing cognition – Application to Resting-State Dataset





# Linking dynamic FC to ongoing cognition – Application to Resting-State Dataset

Section on Functional Imaging Methods

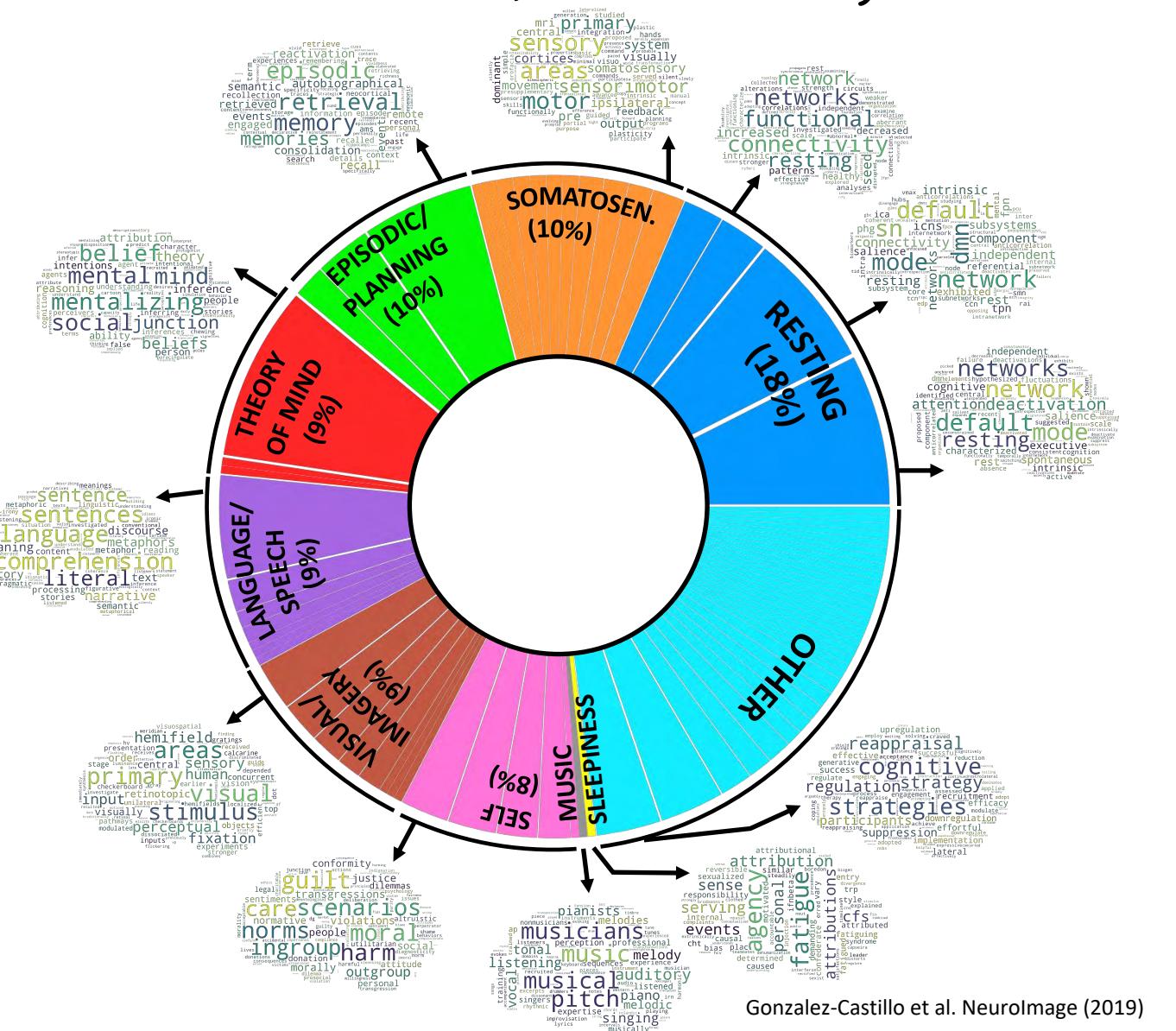
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- \* At the group level, only 78 topics (out of 400) were marked as significant outliers (right tail).
- \* Of those 78, very few appeared as significant in multiple occasions (e.g., different spikes)

## INTERIM CONCLUSIONS

- \* Data-driven estimates of covert cognition agree with previous reports of what are the most common mental activities subjects perform during rest.
- \* "dFC + Deconvolution + Neurosynth" can help us uncover the cognitive correlates of distinct dFC patterns during task and rest.
- \* Dynamic FC is modulated by covert cognition during rest

Top 15 topics most often found as associated with corners in dynamic FC Embeddings





# Linking dynamic FC to ongoing cognition – Problem solved? Umm, not really.

\* We cannot evaluate the accuracy of individual guesses (decoding events)

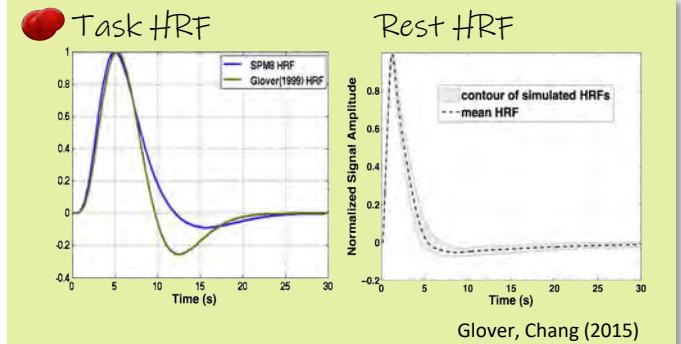
\* Validation / Decoding always through the lens of task-results:

- NeuroSynth is built using results from task-based studies
- Differences in HRF between rest and task
- Uncertainties in interpretation
  - Visual attention vs. Visual imagery
  - Overlap with unconscious brain activity



- Limited specificity of Topics
- Vocabulary biases
- Inability to incorporate priors to contextualize predictions
- Most sampled studies are task-studies

Rubin et al. (2017); Yarkoni et al. (2011)

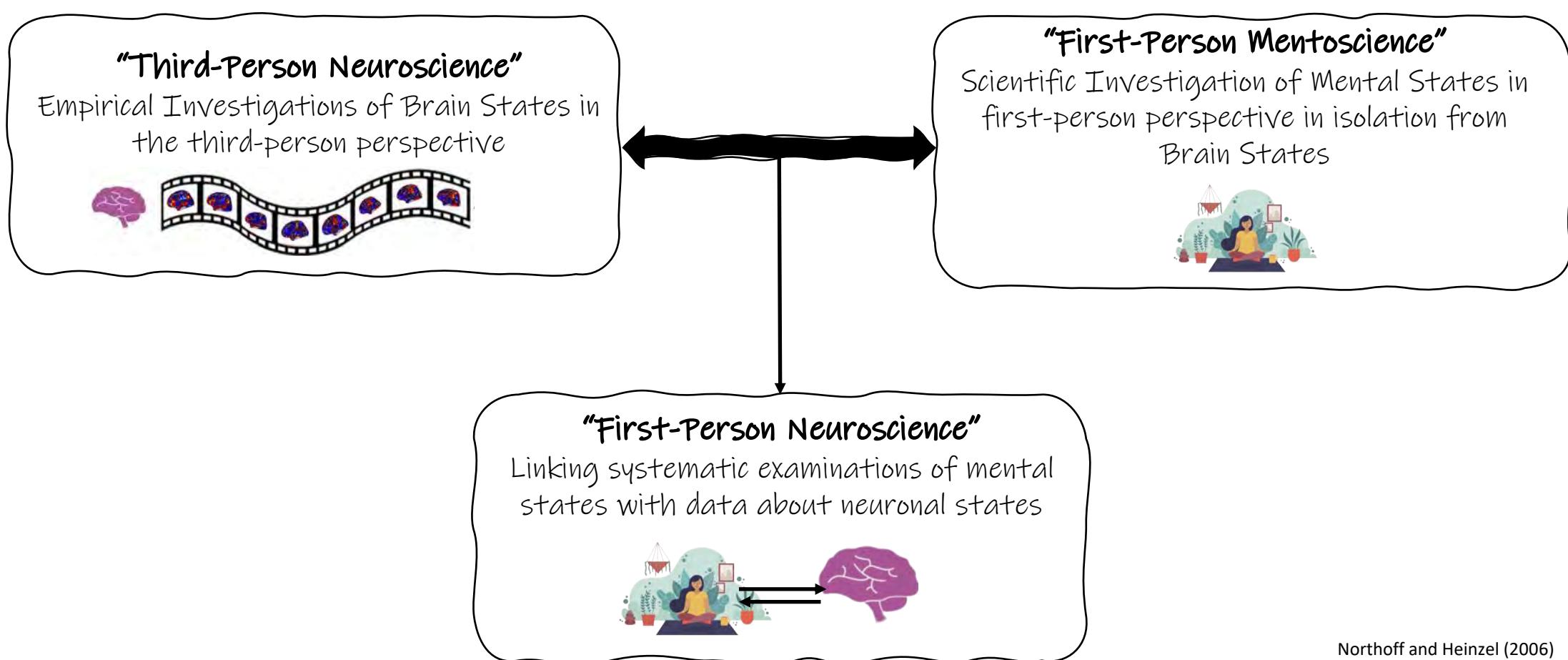
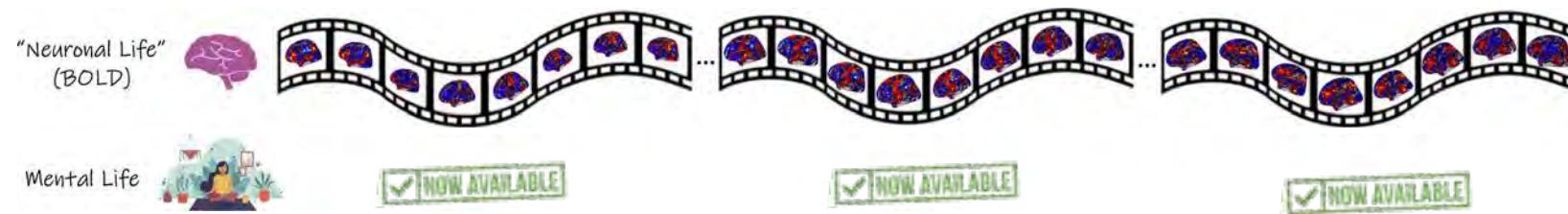


\* Best-case scenario, we likely only captured most prominent "events/periods":

- Missed a lot of important information needed to fully understand the joint dynamics of brain and "mind" during rest.

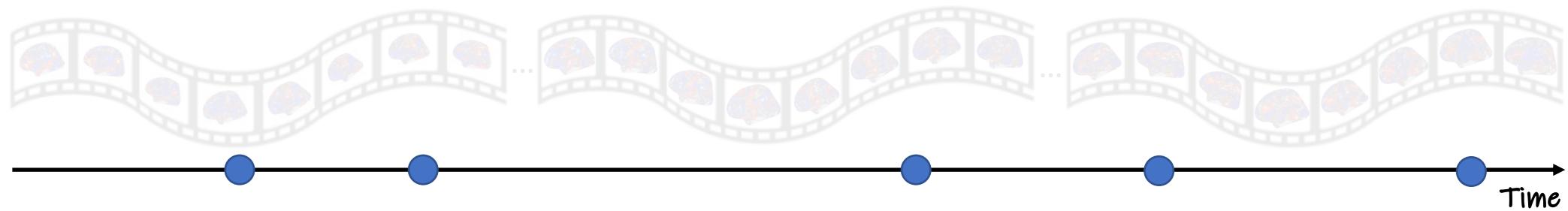


# Linking dynamic FC to ongoing cognition – Future Steps





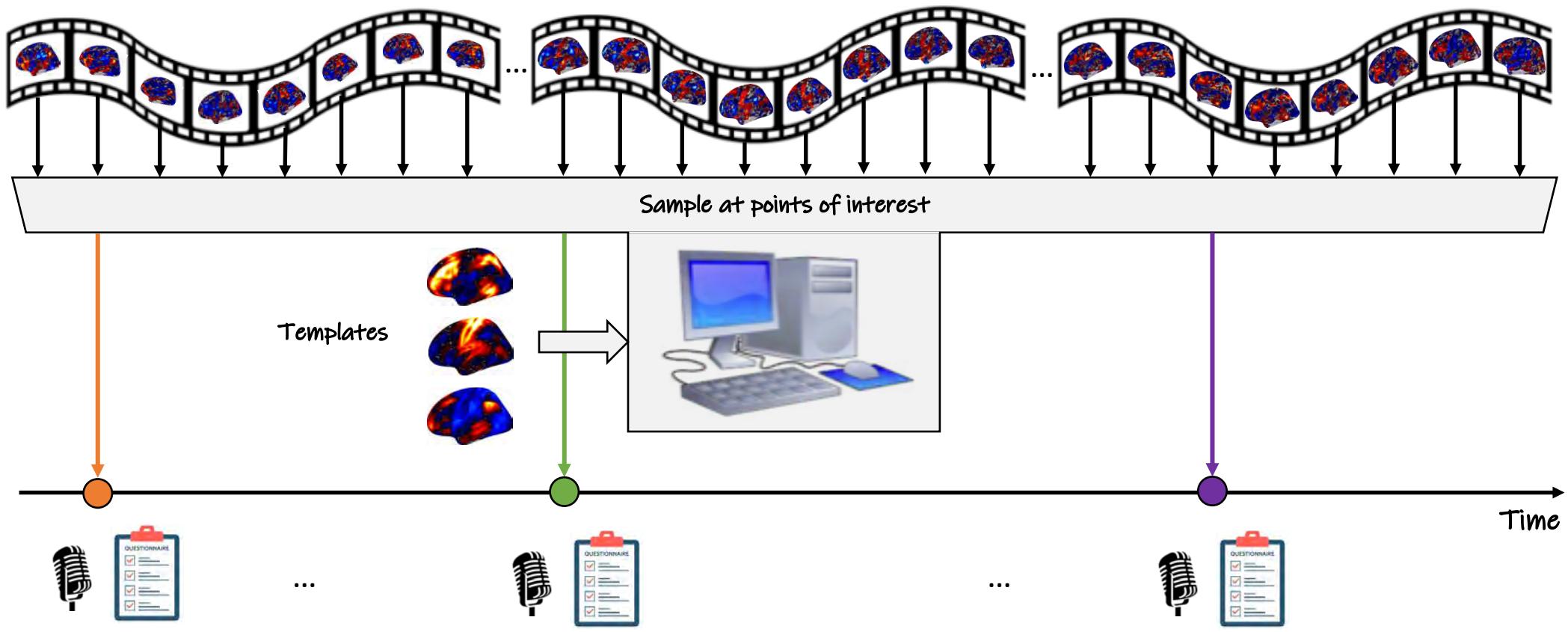
## Linking dynamic FC to ongoing cognition – Future Steps



Sample at random periods



# Linking dynamic FC to ongoing cognition – Future Steps



Obtain an unconstrained description of mental life [ zero assumptions hard to analyze]

Obtain description of mental life across pre-determined dimensions [ assumptions easy to analyze]



Resting State  
Functional Connectivity

=>

Mean Effect  
(Static Functional Connectivity)

&

Small Fluctuations over the Mean  
(Dynamic Functional Connectivity)

\* Since one of the original reports (Chang et al. 2010), we have made substantial progress:

- The avid debate regarding its significance (artifactual or neuronally meaningful) is starting to settle.
- Covert ongoing cognition is a modulating factor of Dynamic FC estimates.
- Reports of significant differences in dynamic FC across populations (not captured by static measures).

\* Yet, there is substantial work ahead of us:

- What are the relative contributions of the different sources of variability to Dynamic FC?
- Linking brain states (dFC) and mental states will require taking into consideration first-person reports





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