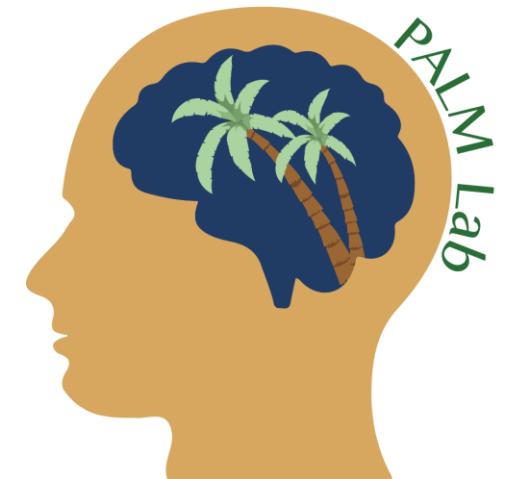


# Multivariate classification shows associative learning reduces working memory load

William X. Q. Ngiam

School of Psychology, Adelaide University



# A quick introduction to me

I study **attention and working memory** – how information is selected and held in mind for ongoing perception and cognition.

# A quick introduction to me

I study **attention and working memory** – how information is selected and held in mind for ongoing perception and cognition.

One key feature of this system is that it is **capacity-limited**. Measuring the capacity limits of visual attention and working memory requires understanding how information is **represented** in working memory.

# The point of this talk

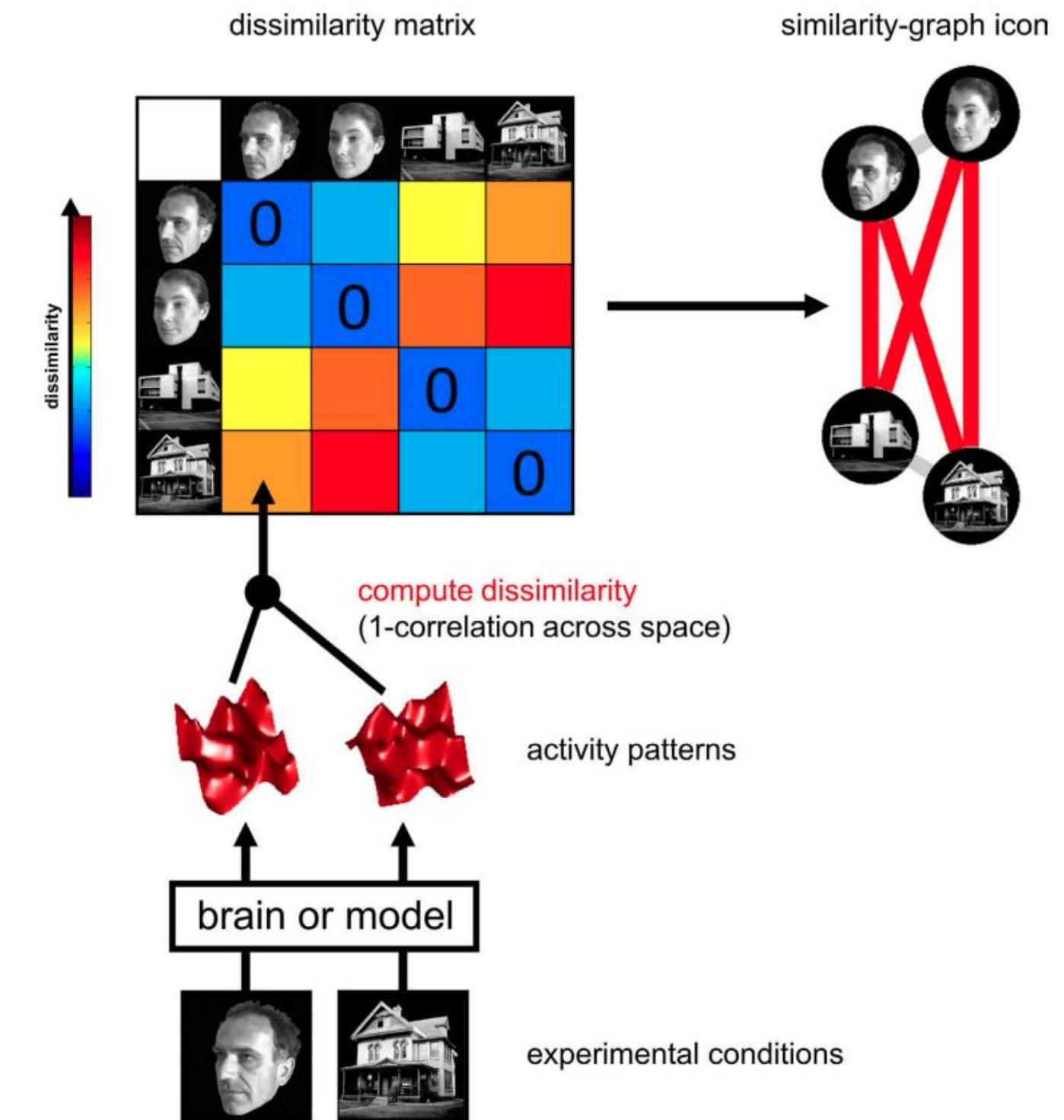
Cognitive theories ought to make predictions about neural signals and decoding results, such as those from **representational similarity analysis**.

A formal modeling approach that **incorporates representation as part of the cognitive model** might help make theory-driven predictions about neural representations.

# Representations to neuroscientists

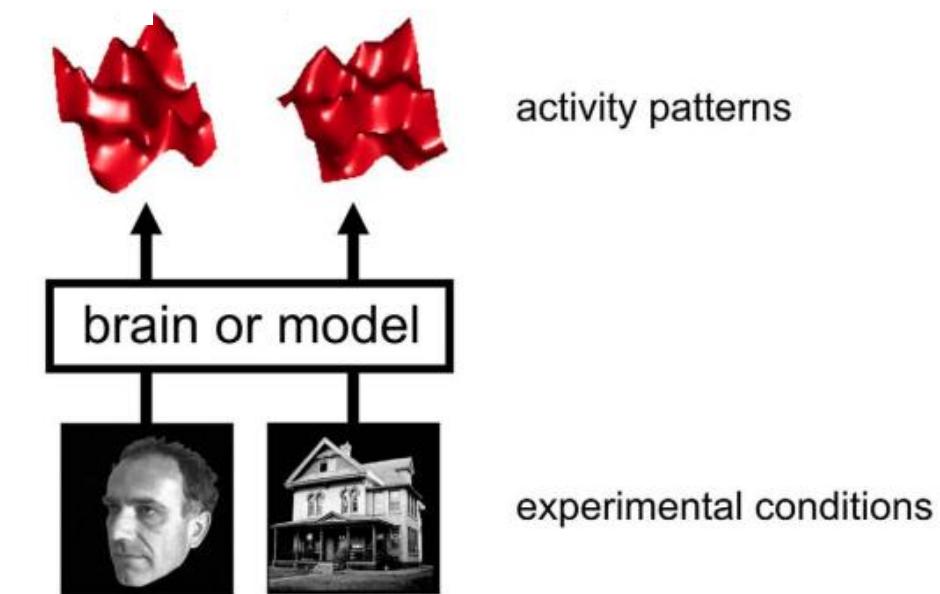
Other neuroscientists are using training neural network models to achieve human-like performance or using machine learning to decode neural representations.

One such method is **representational similarity analysis**:

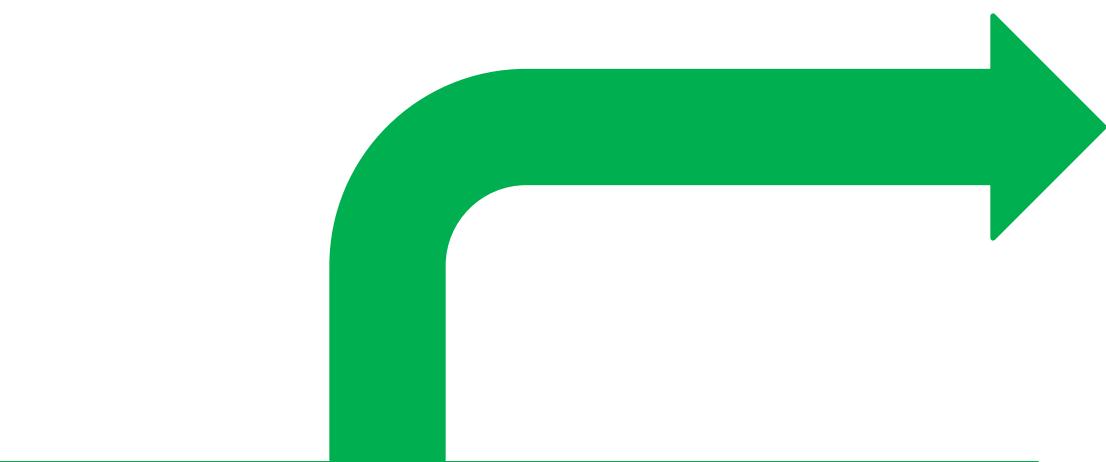


# Representational similarity analysis

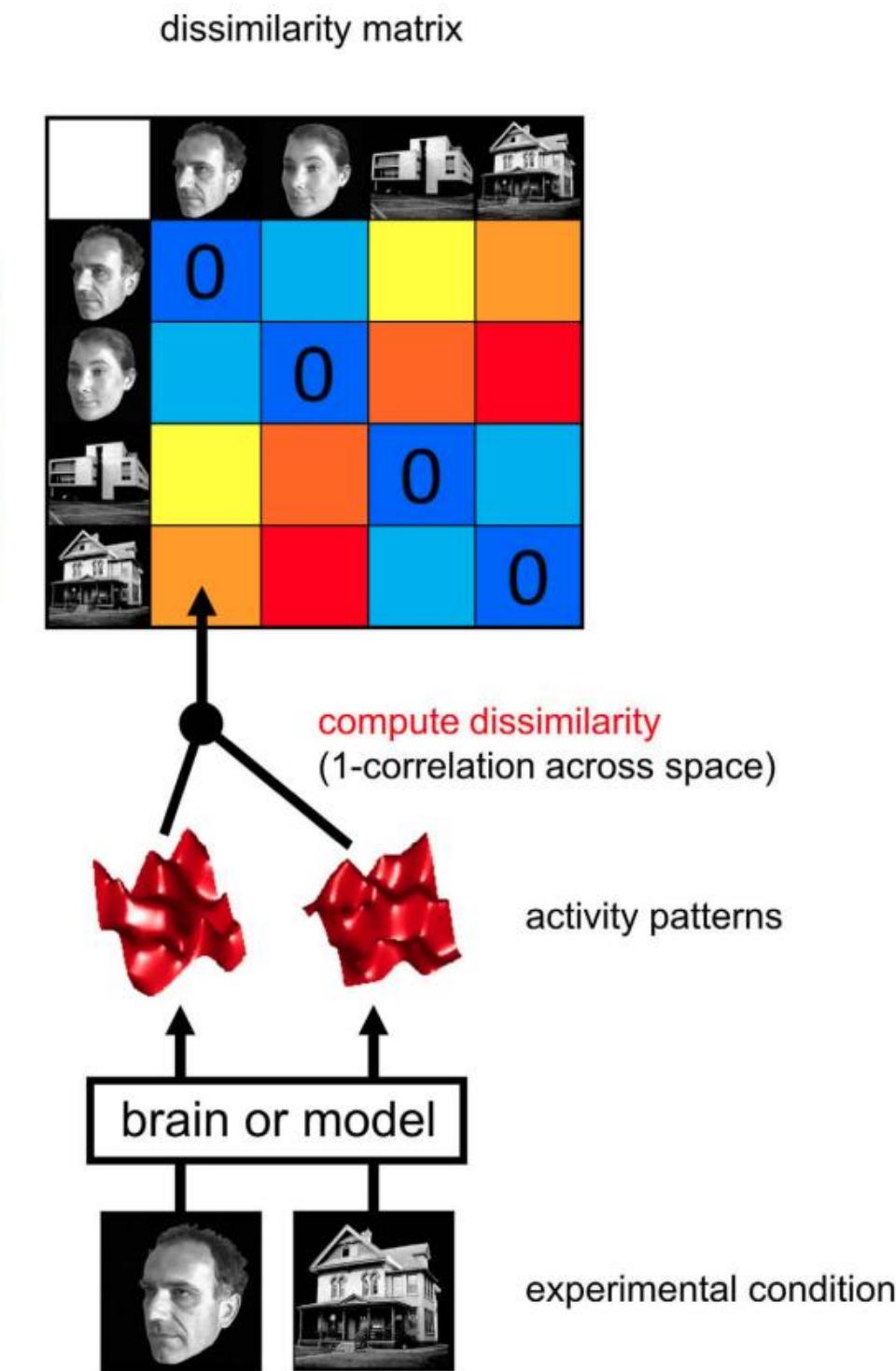
The researcher collects neuroimaging data under experimental conditions



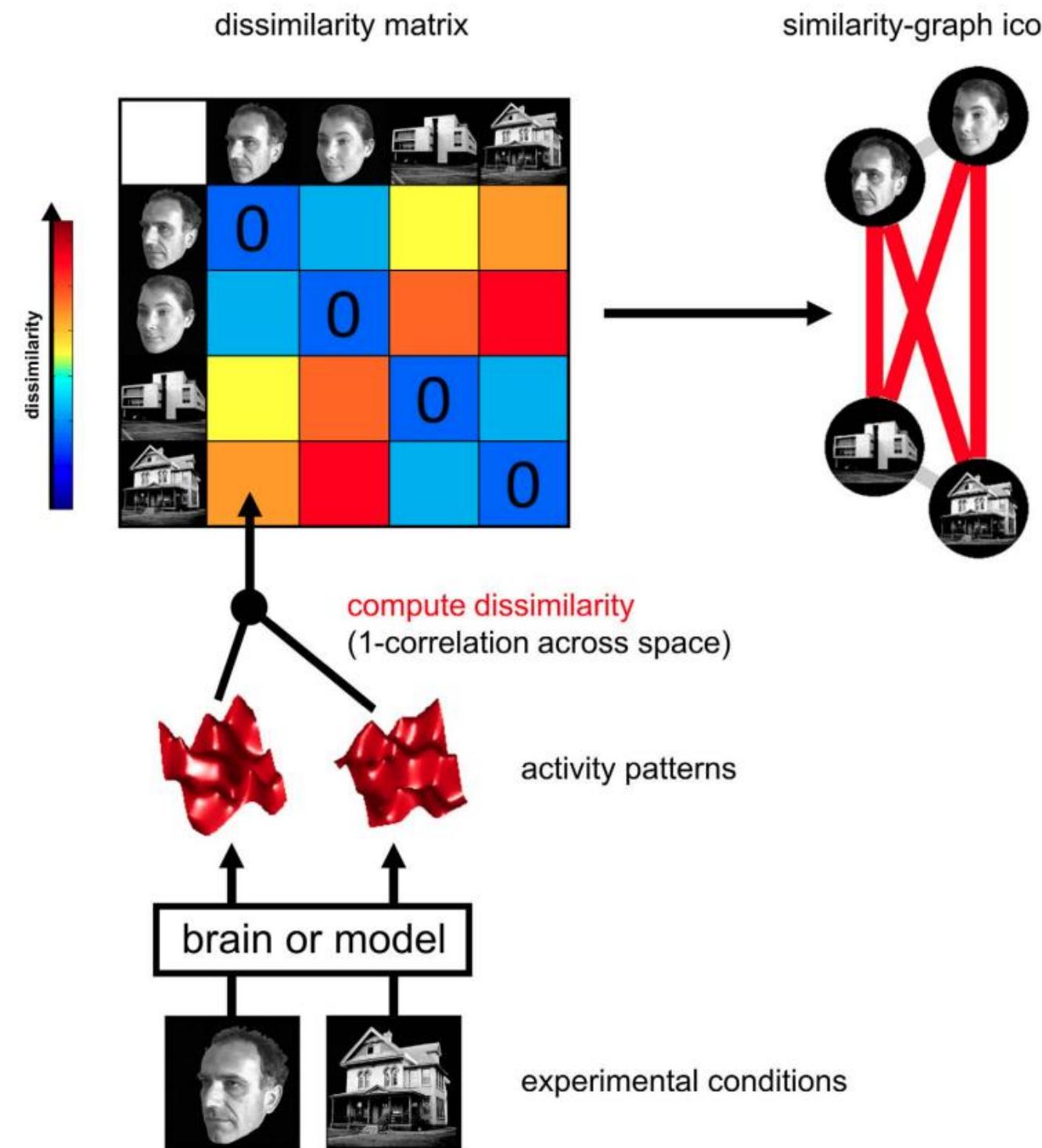
# Representational similarity analysis



The researcher uses machine learning to decode the conditions and computes the distance between conditions

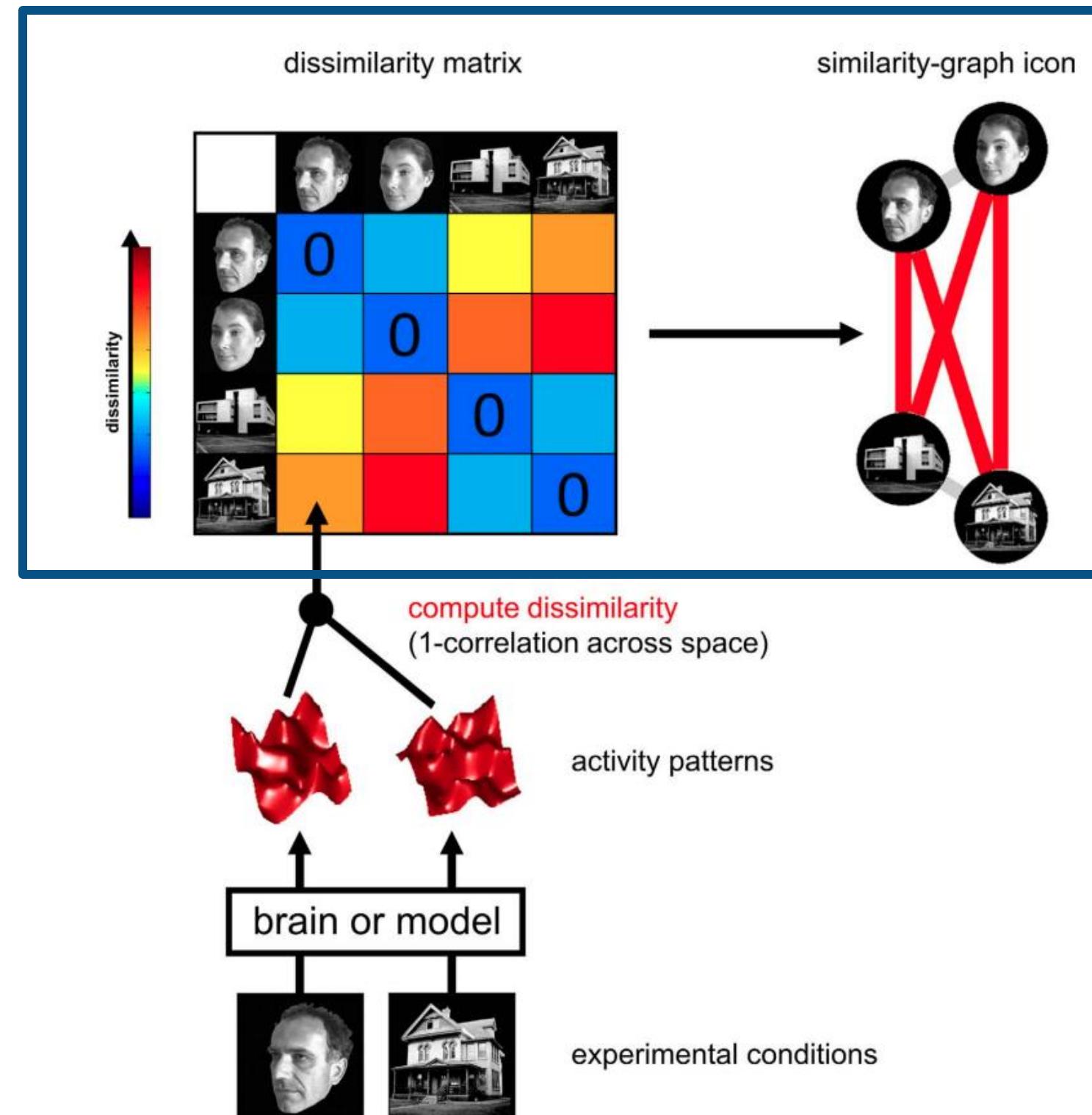


# Representational similarity analysis



The dissimilarity can be visualised using multidimensional scaling or other methods.

# Representational similarity analysis

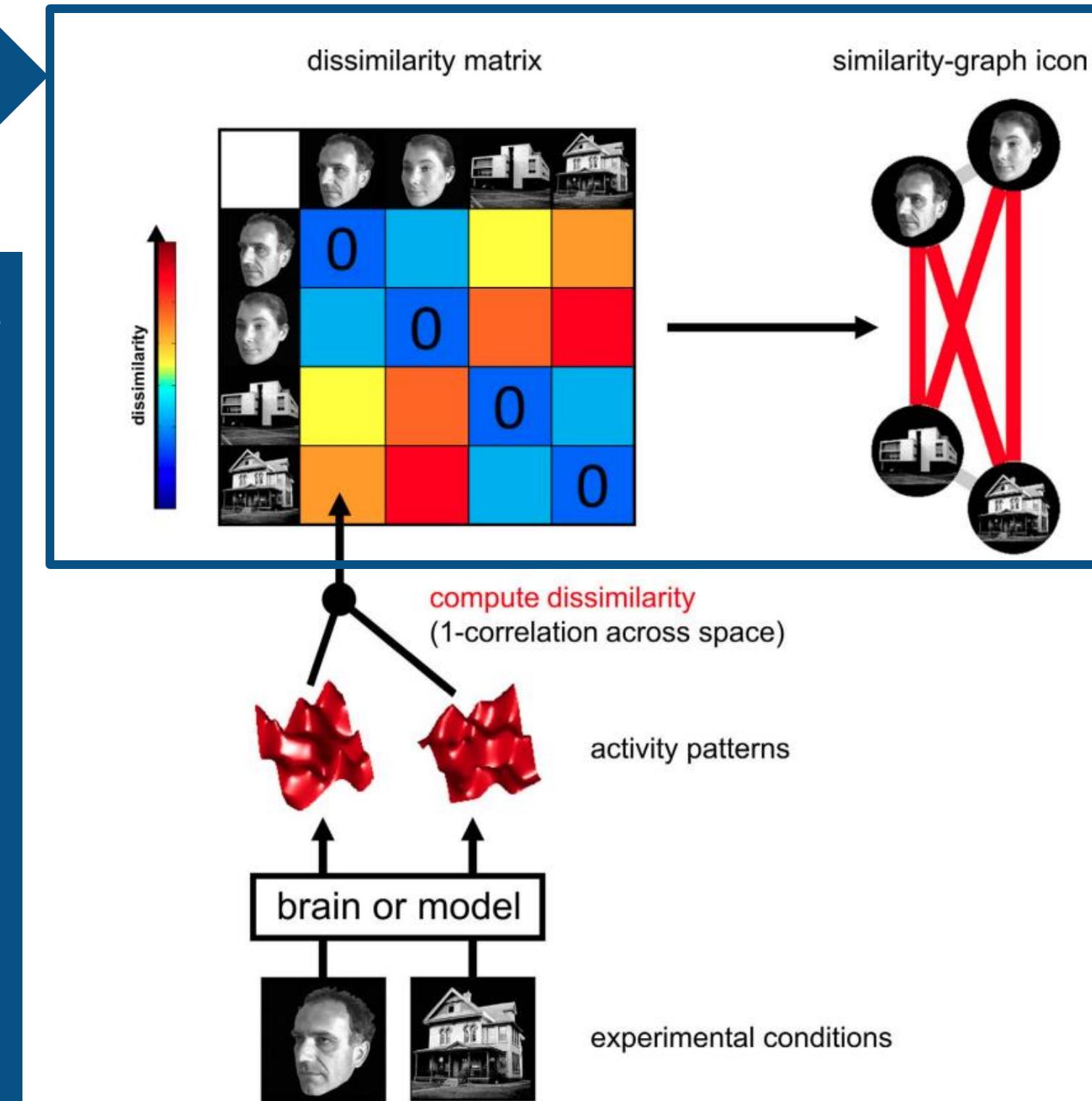
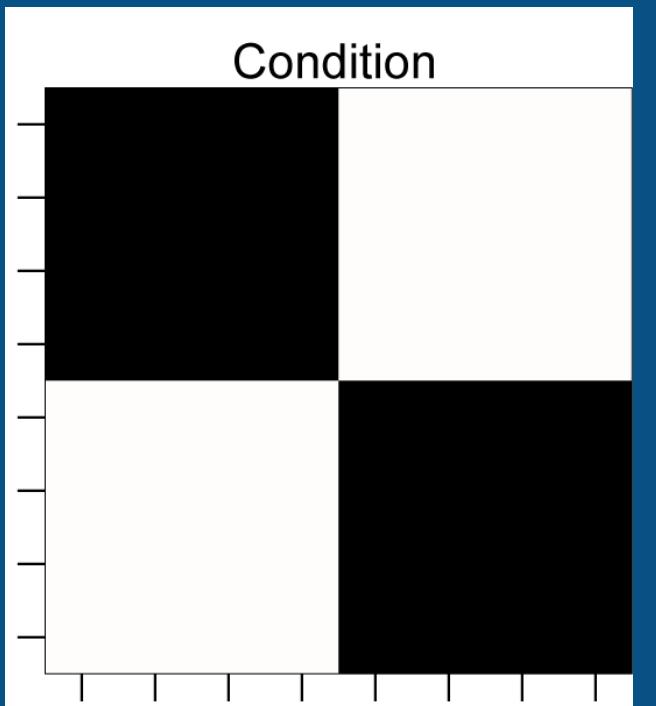


Do these results provide  
a veridical basis for  
models of cognition?

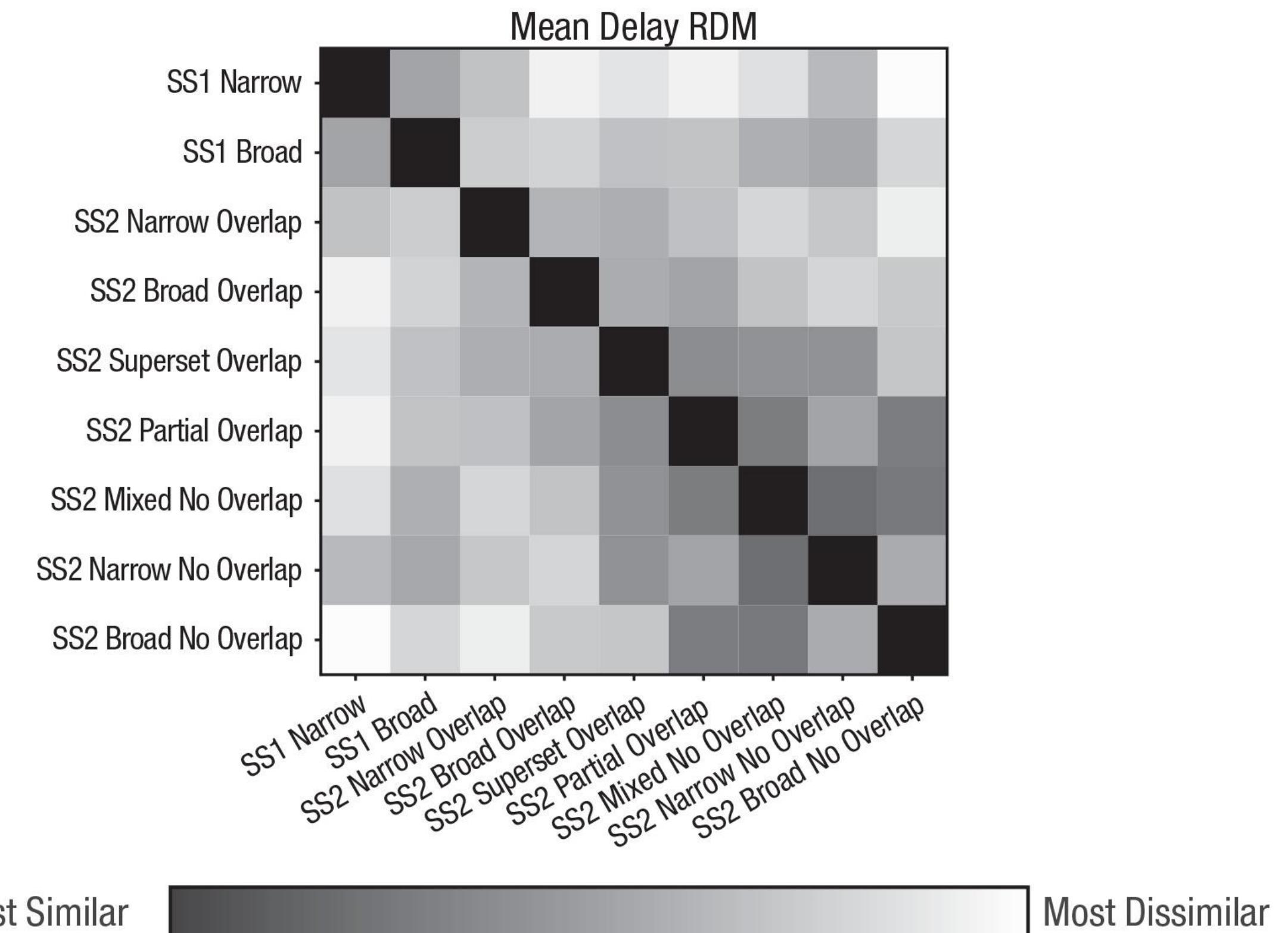
We might be **HARKing**  
and/or drawing ad-hoc  
conclusions.

# Representational similarity analysis

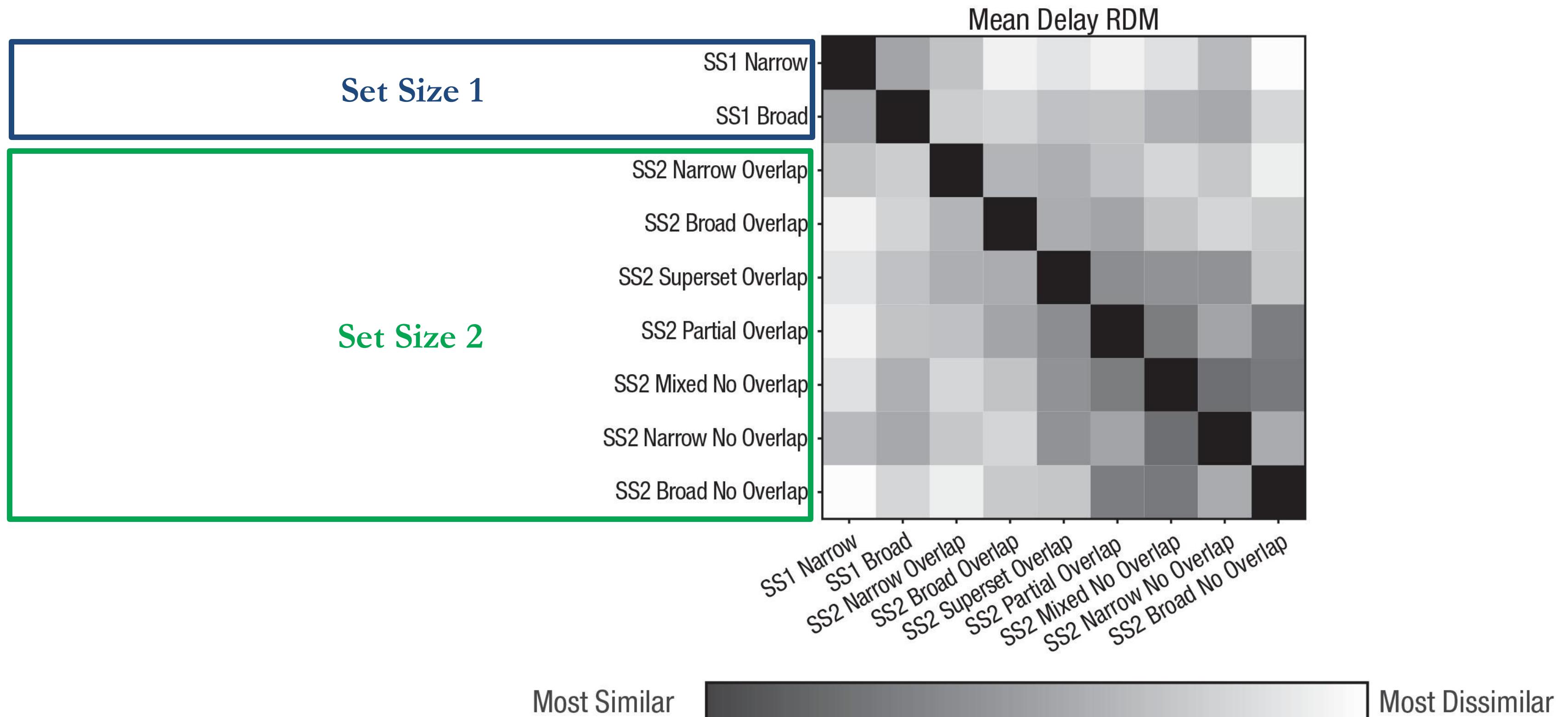
We might create models to predict the observed representational (dis)similarity



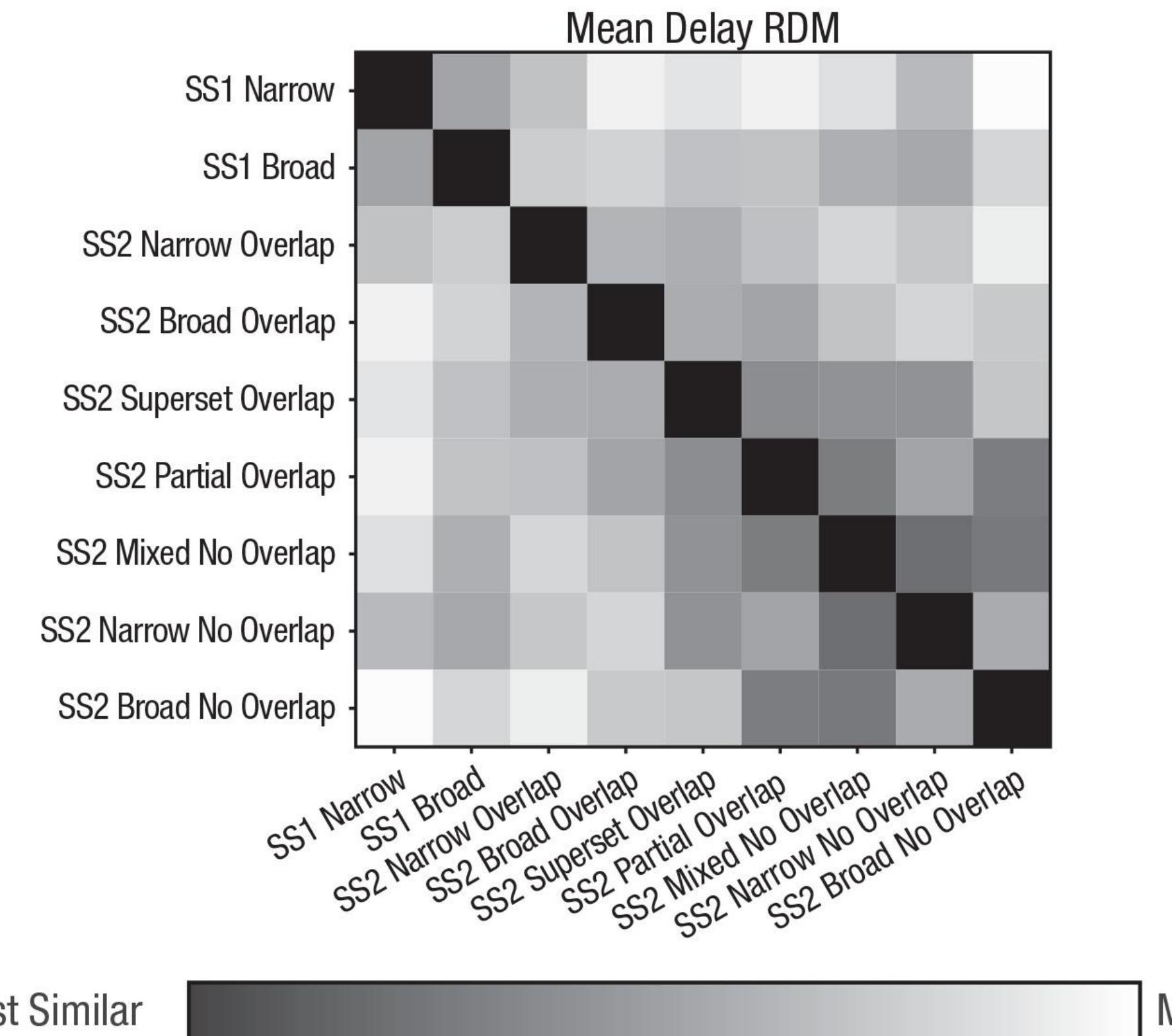
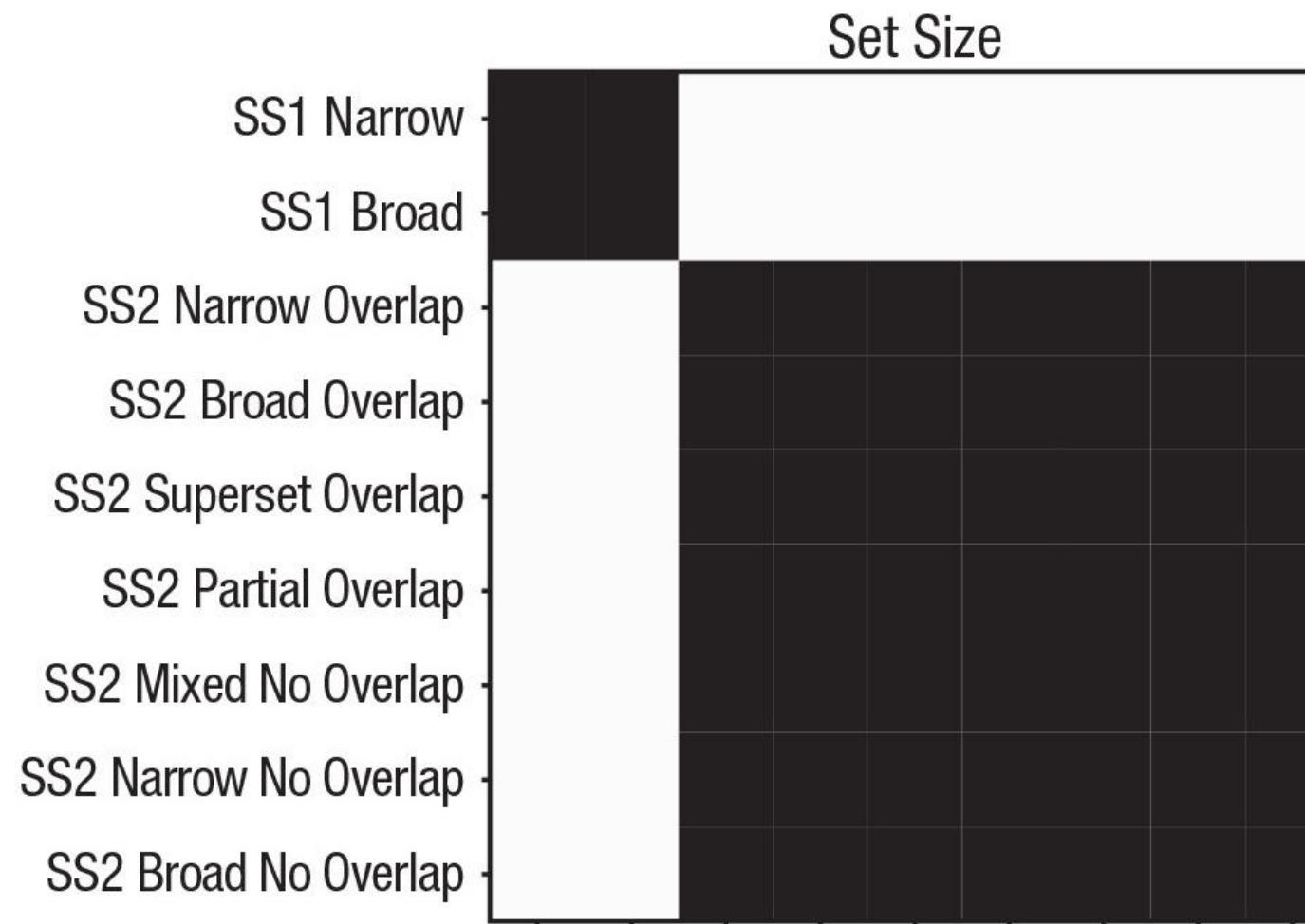
# Decoding load in working memory



# Decoding load in working memory



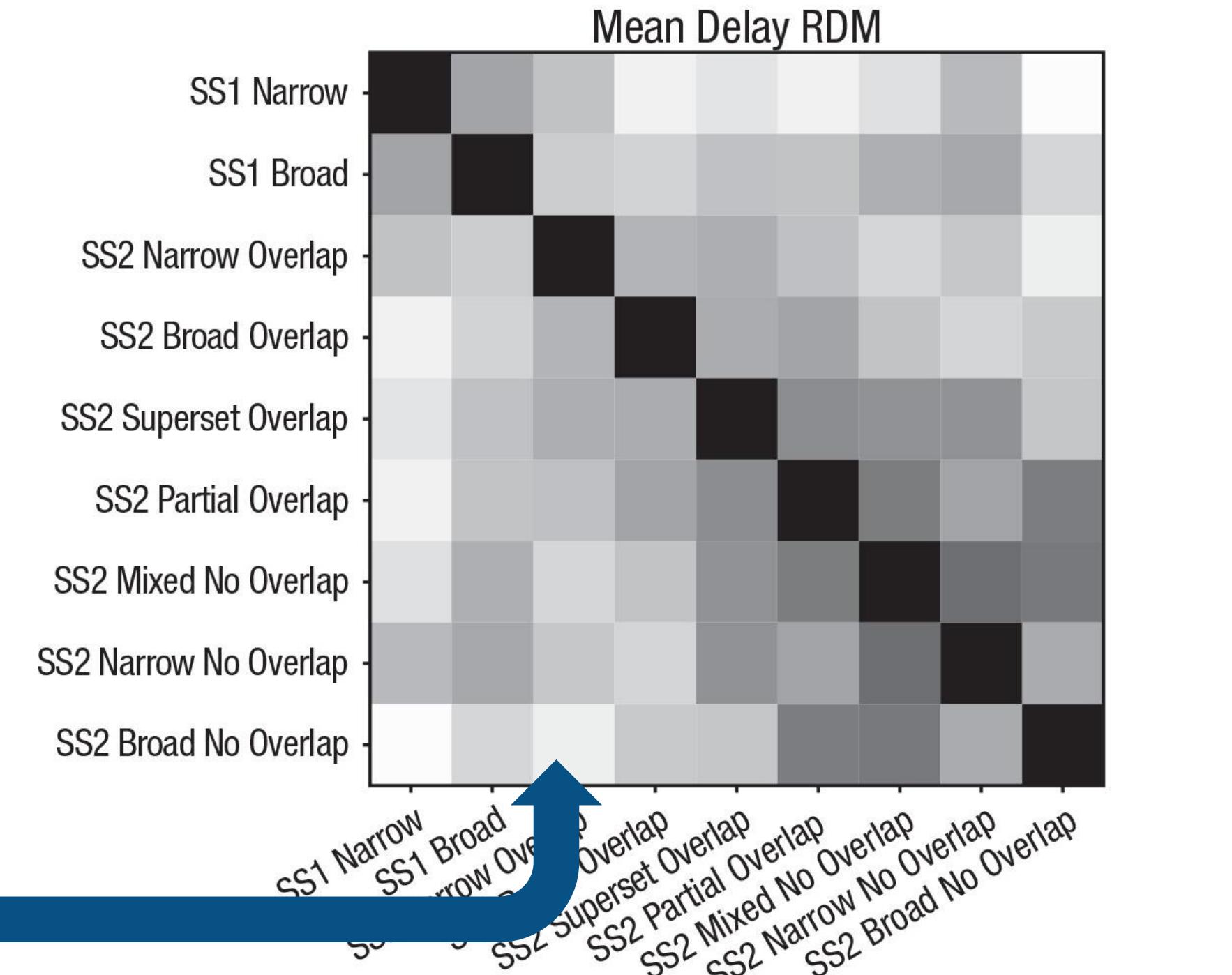
# Decoding load in working memory



# Decoding load in working memory



A “model” to predict neural  
representational (dis)similarity



Most Similar

Most Dissimilar

# Decoding load in working memory

We were interested in how associative learning influences working memory operations.

One proposed operation is that associative learning leads to “chunking” processes – representations where separate items are bound into a “chunk”.

# Training

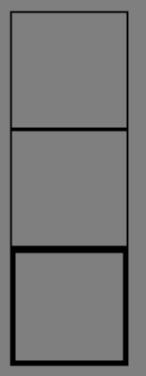
Trained subjects to learn three color triplets



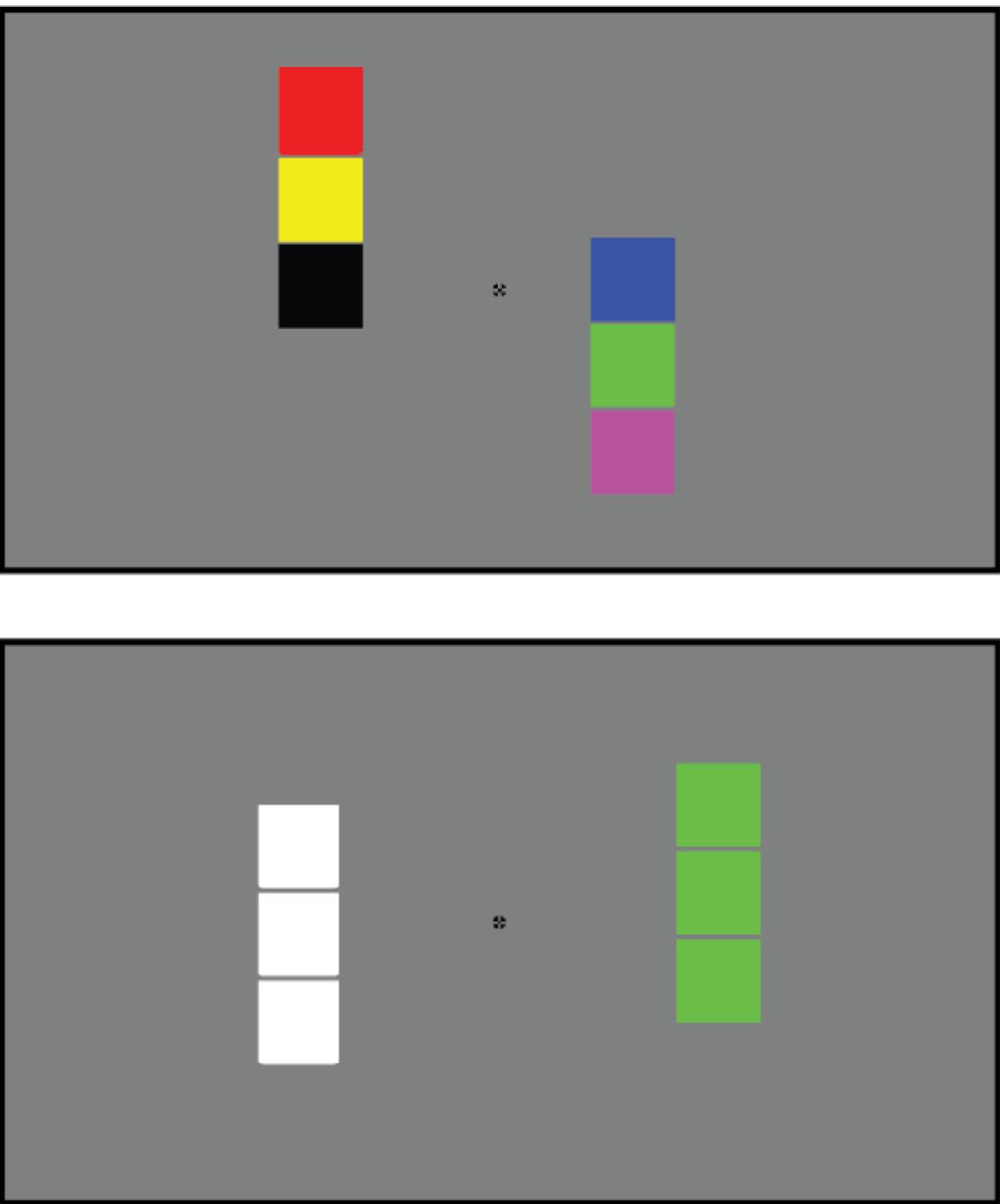
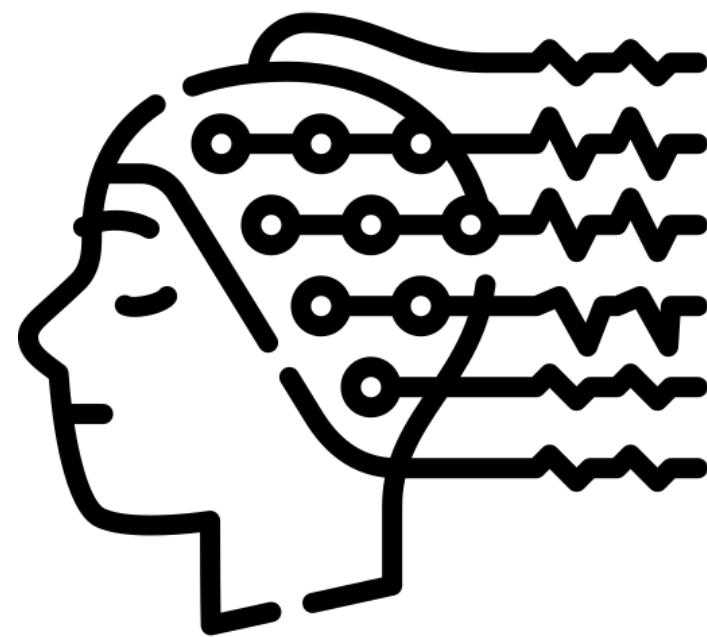




■ . ■



# EEG



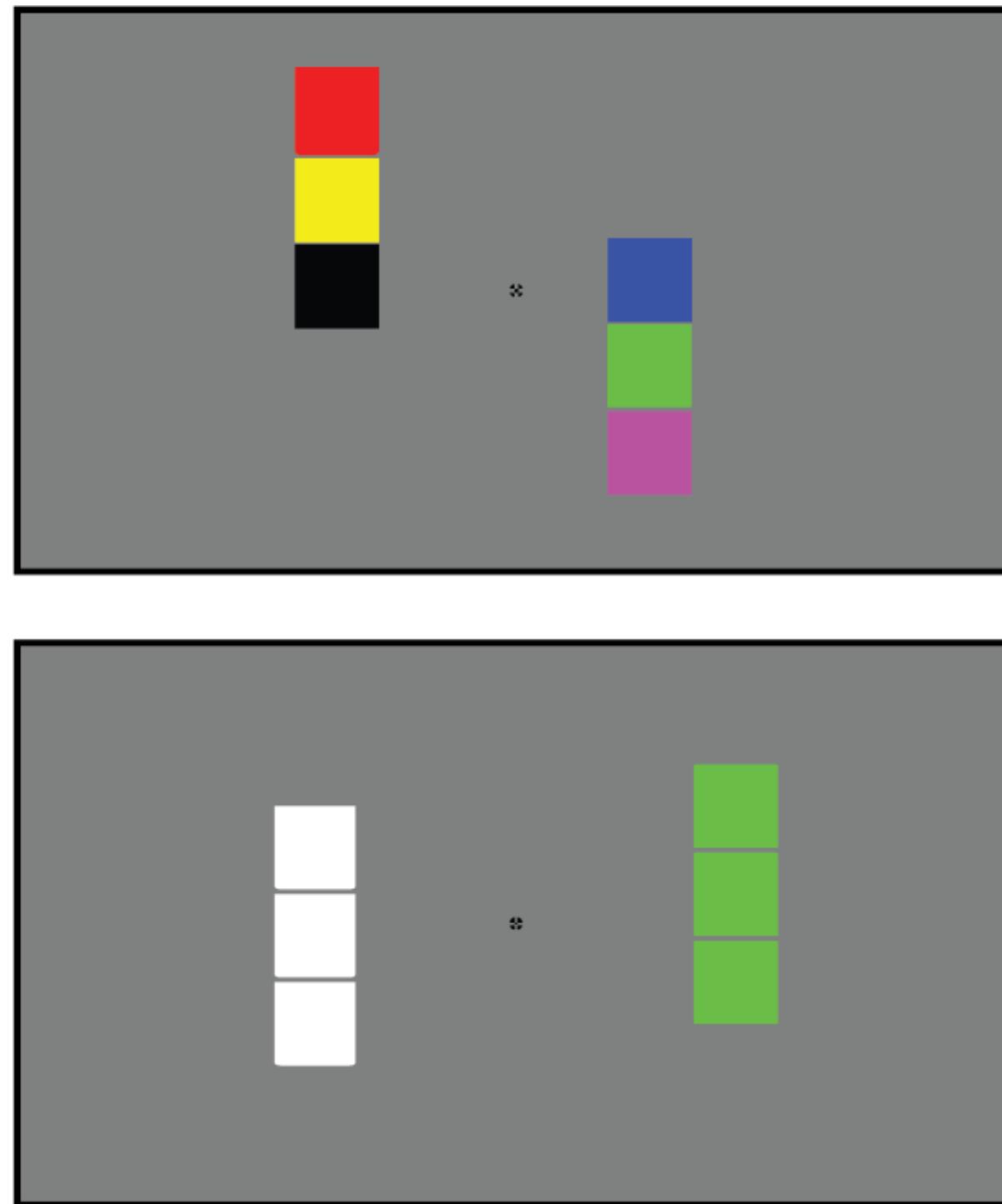
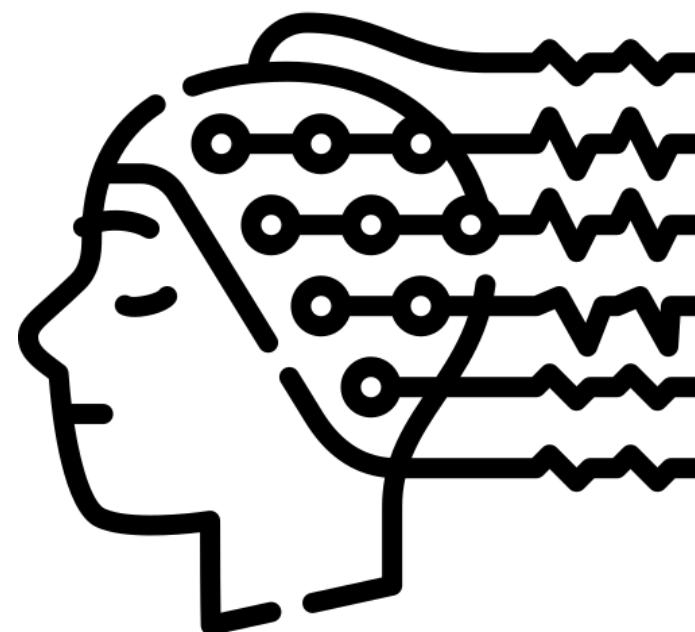
Six random  
Six chunked

Two random

Perceptually equivalent



# EEG



Six random  
Six chunked

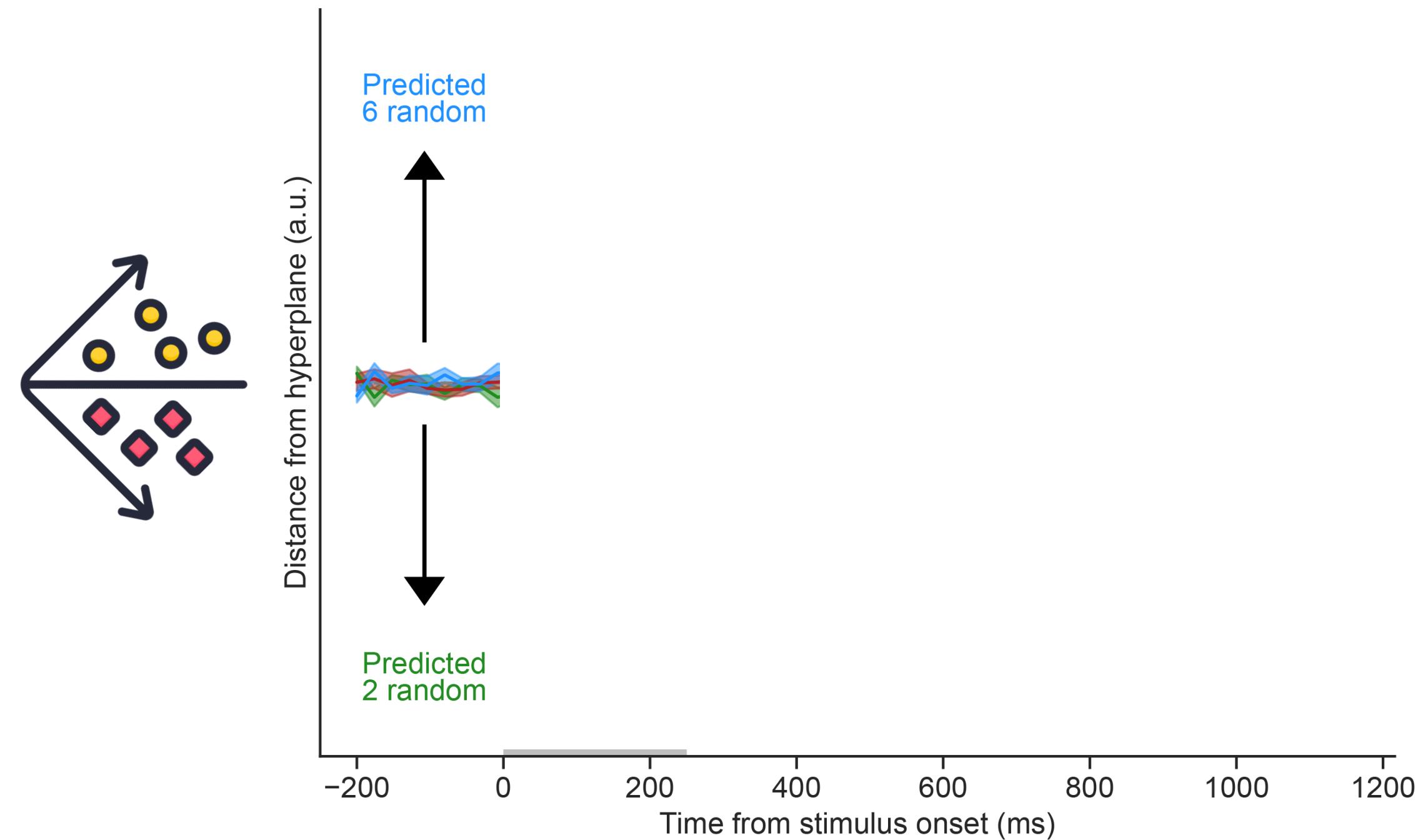
Two random

Perceptually equivalent

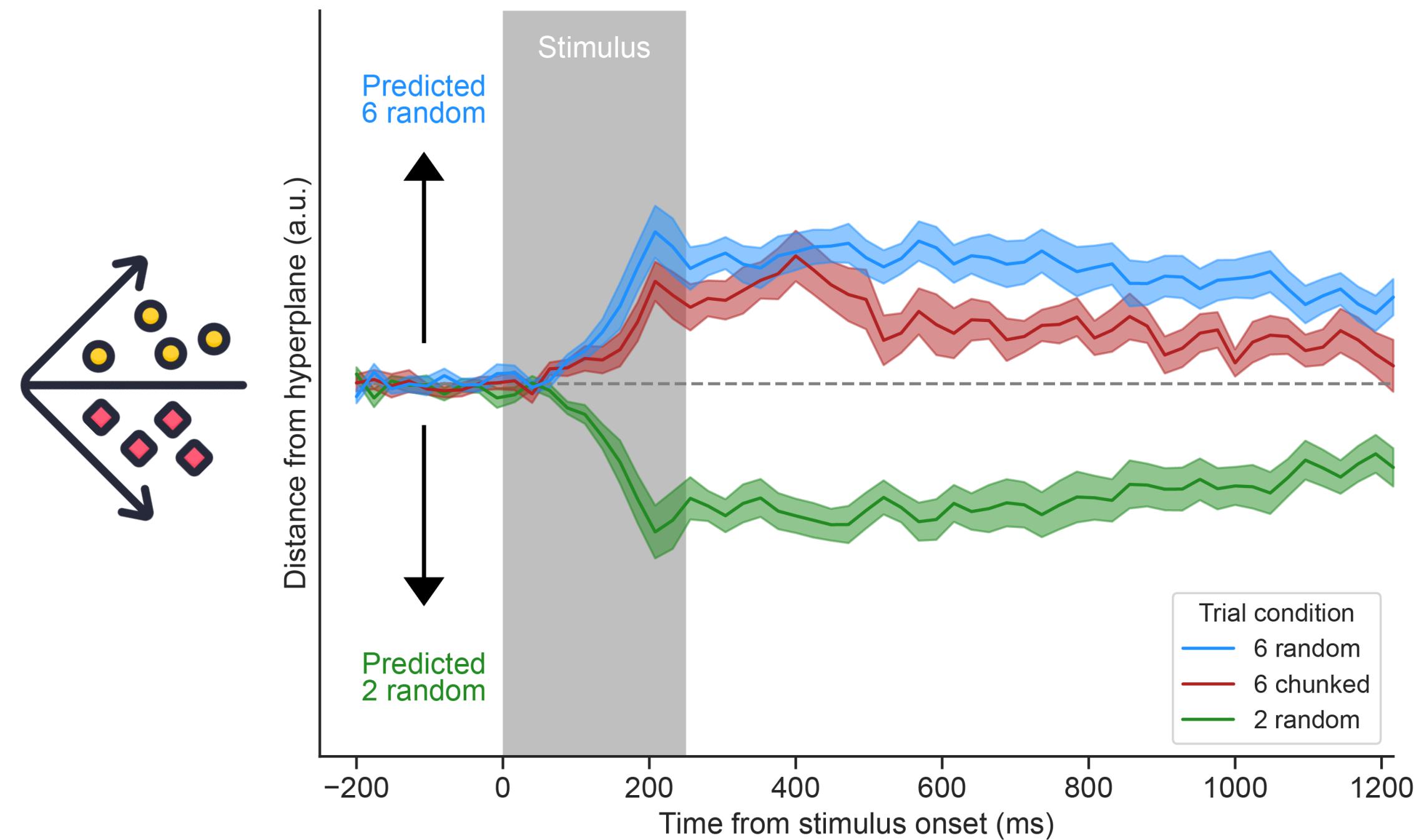


Expectation: “chunking” results in a reduction of item-based load that should be reflected in neural representations

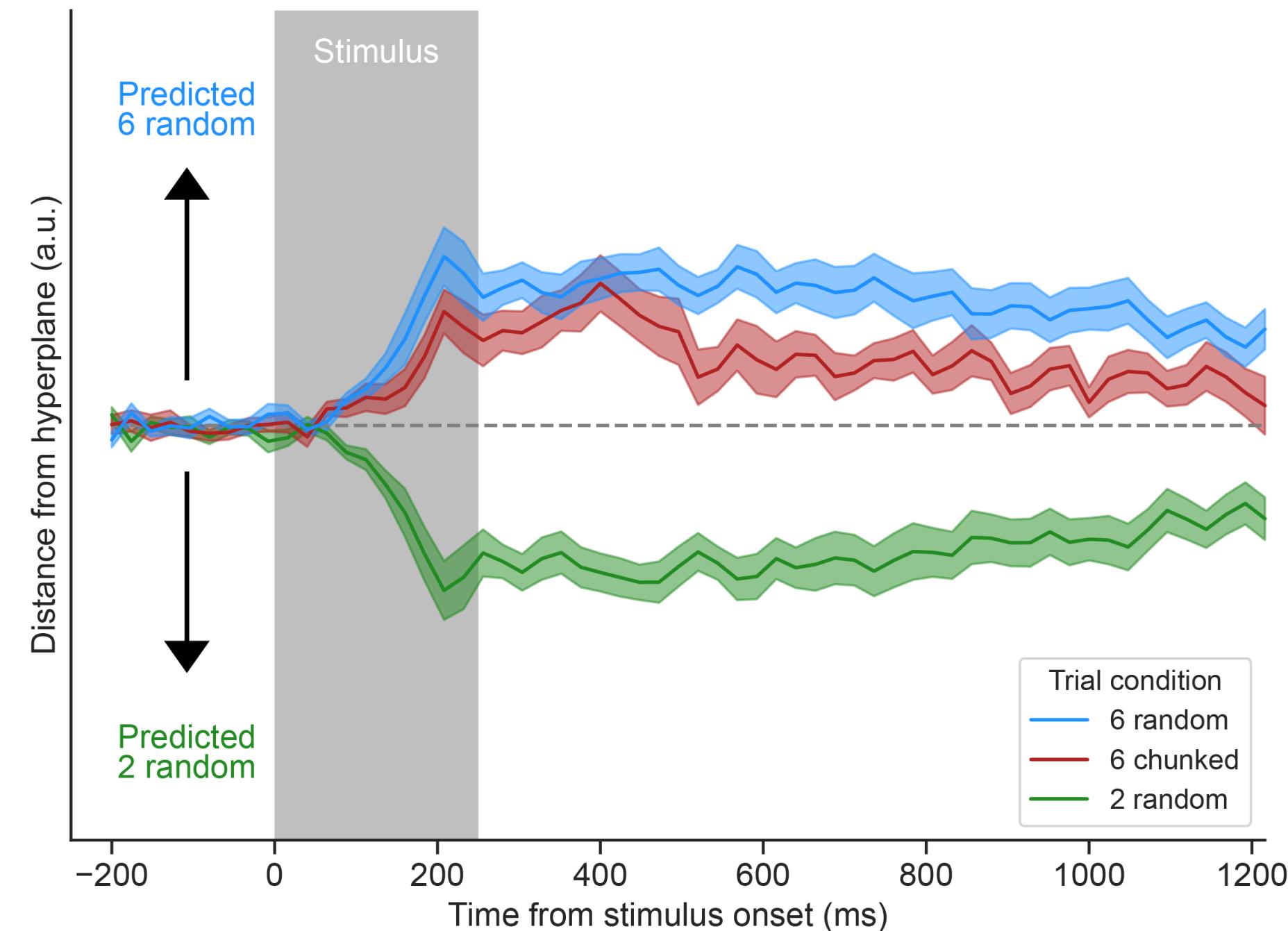
# Train 6 random versus 2 random, test 6 chunked



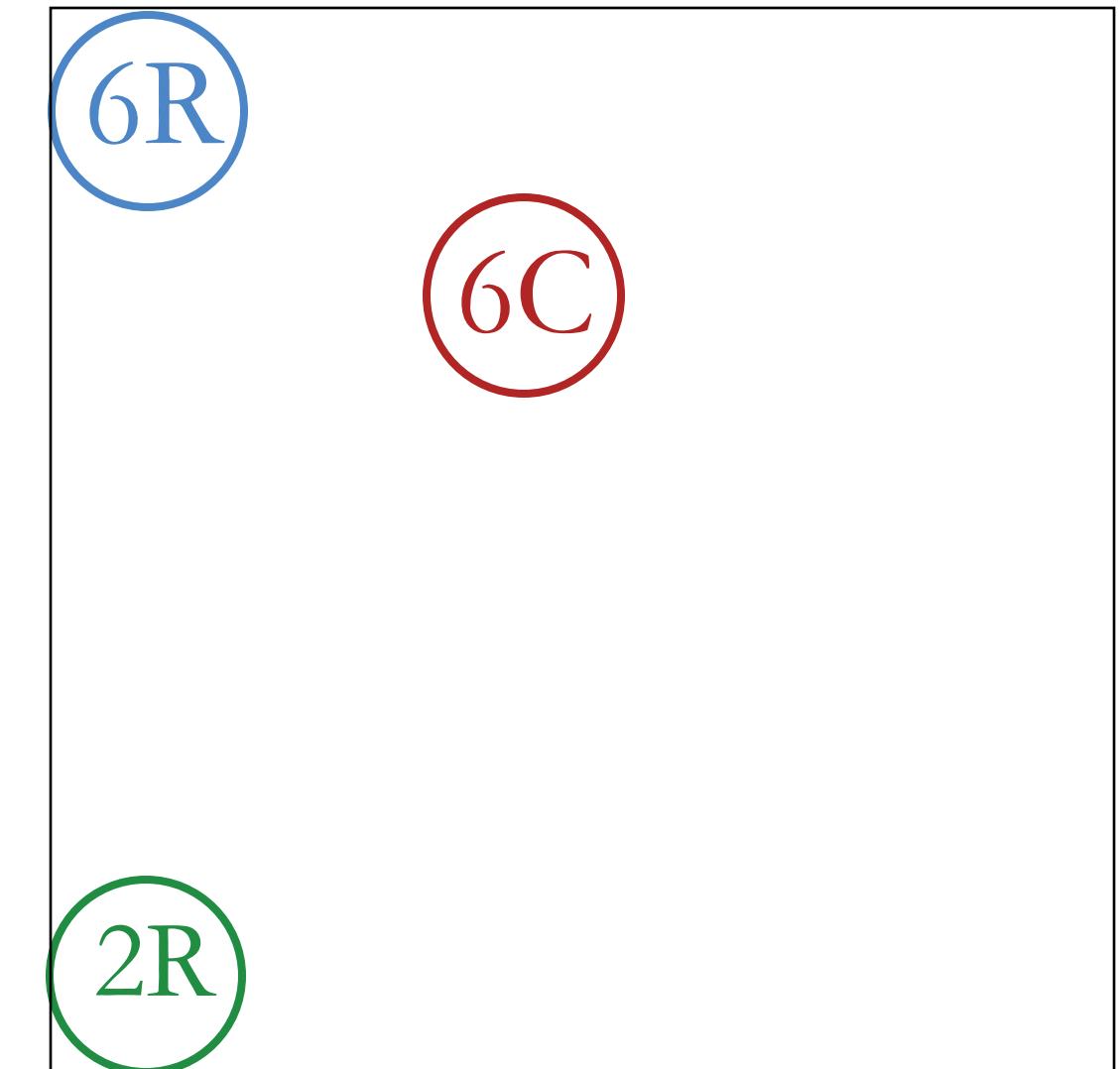
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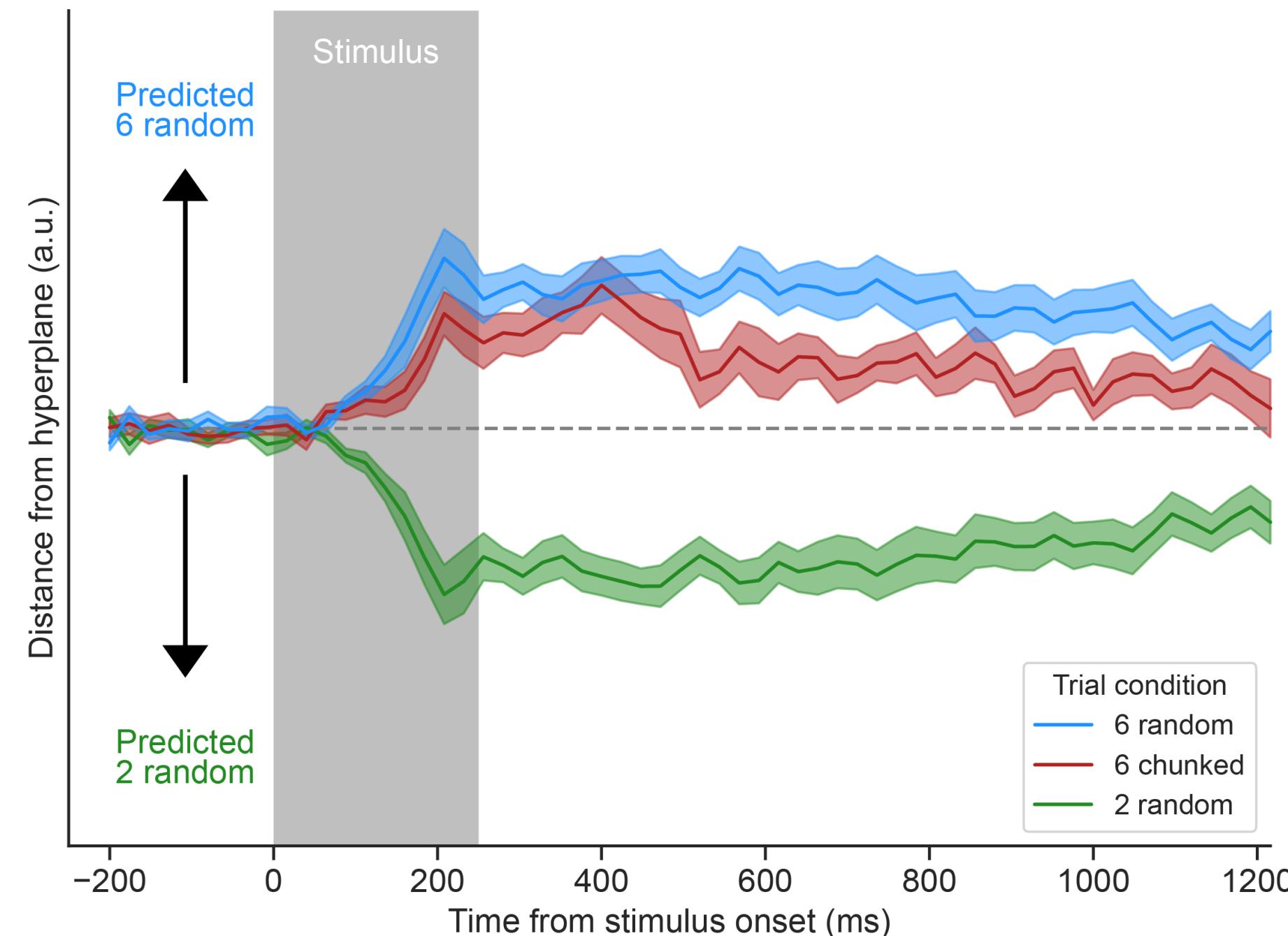
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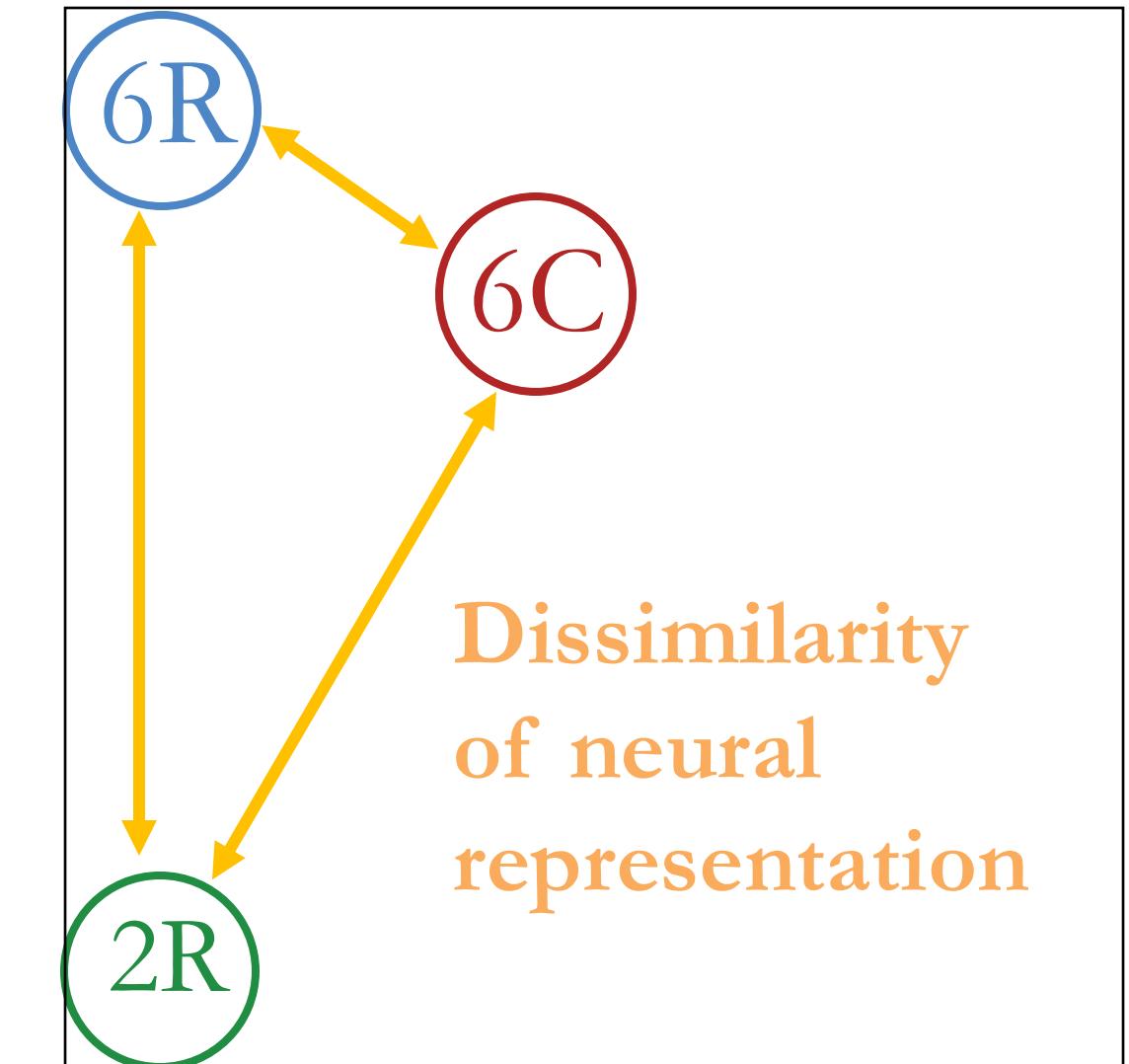
Multidimensional scaling



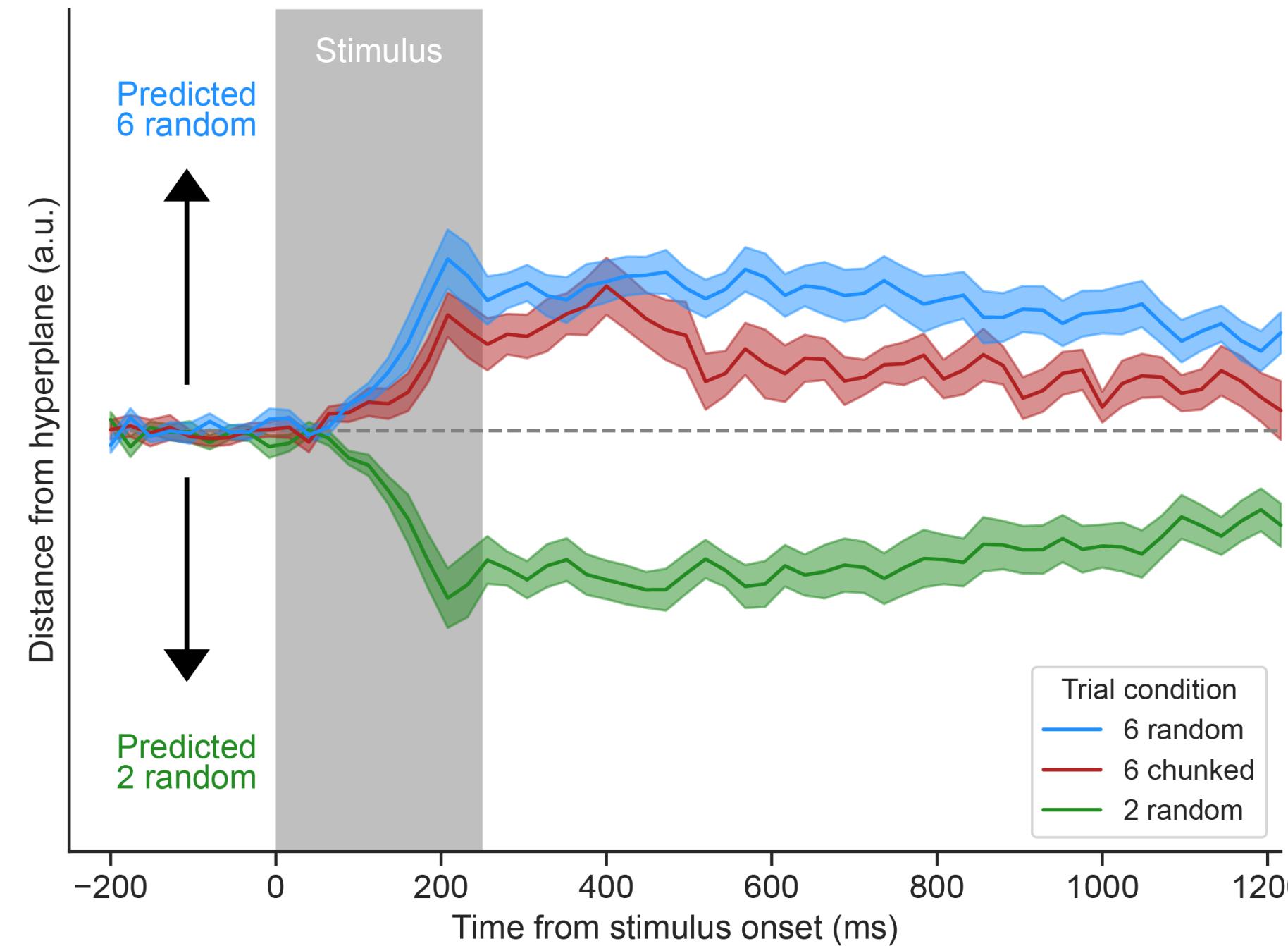
# Train 6 random versus 2 random, test 6 chunked



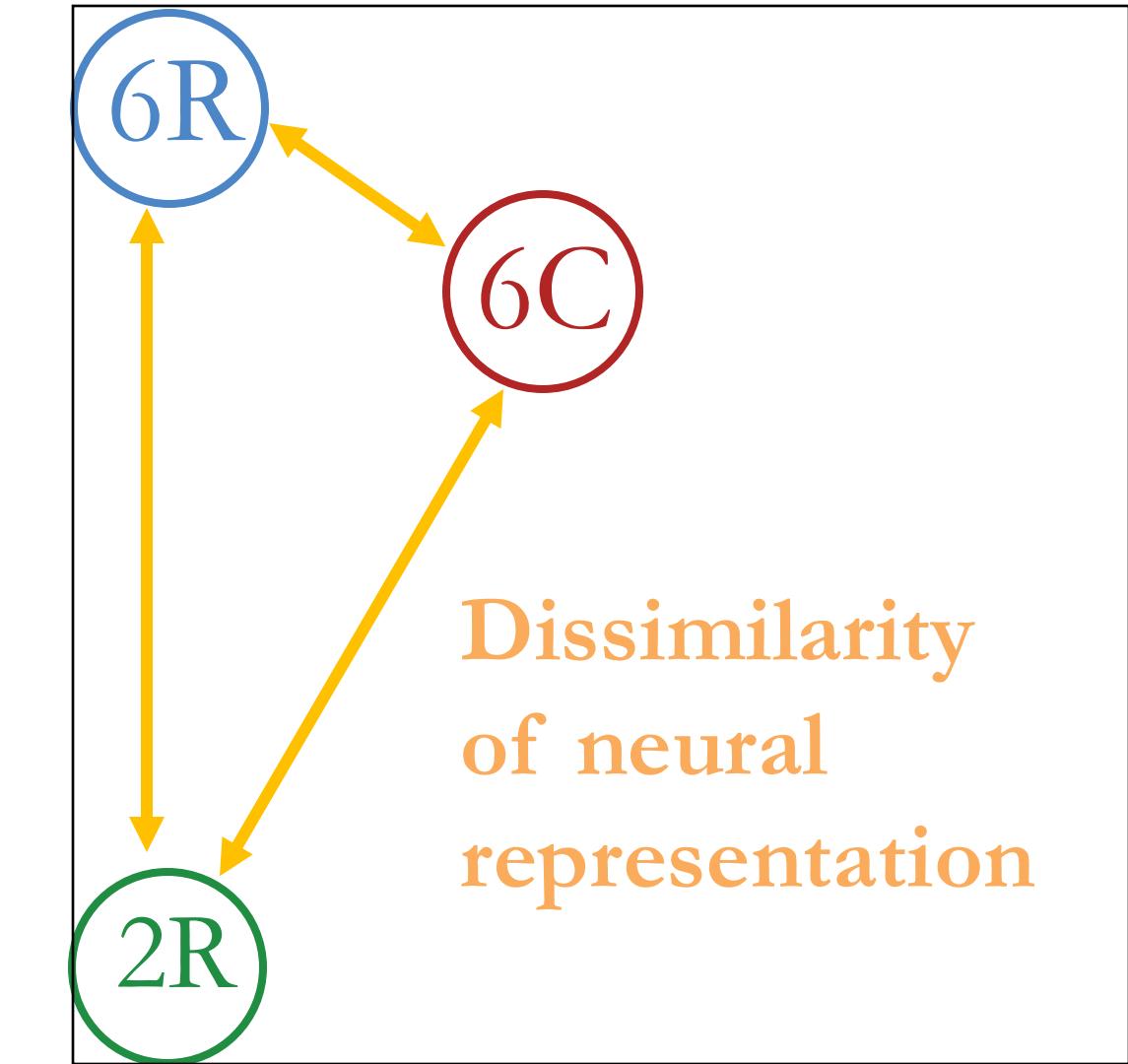
Multidimensional scaling



# Train 6 random versus 2 random, test 6 chunked

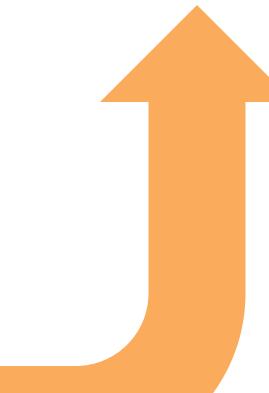


Multidimensional scaling

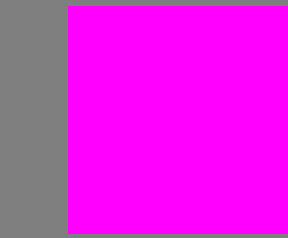
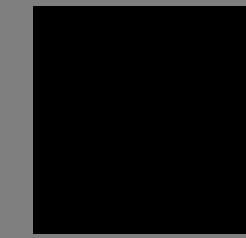
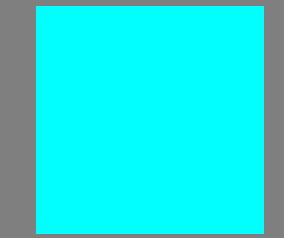
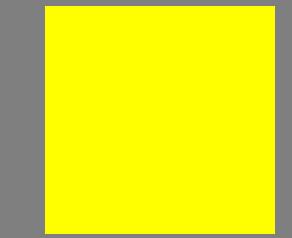
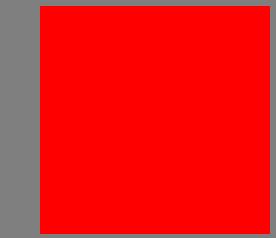
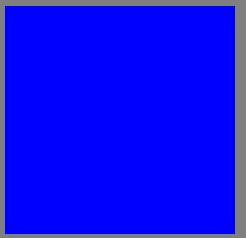


What can we infer from these results?

What cognitive model would predict or could explain this pattern of representational similarity?

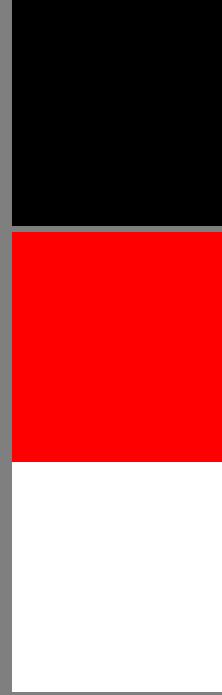
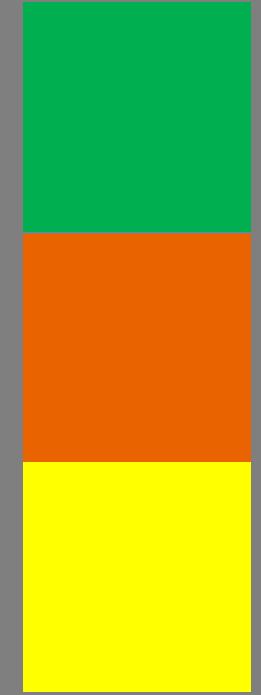
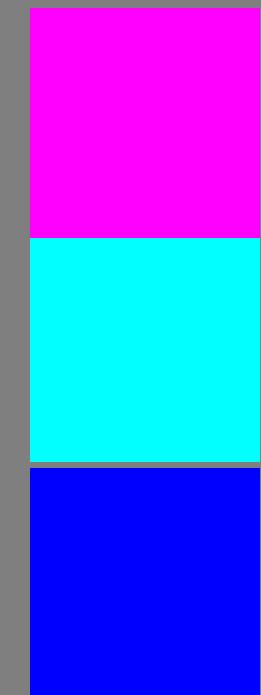


# Awareness Test

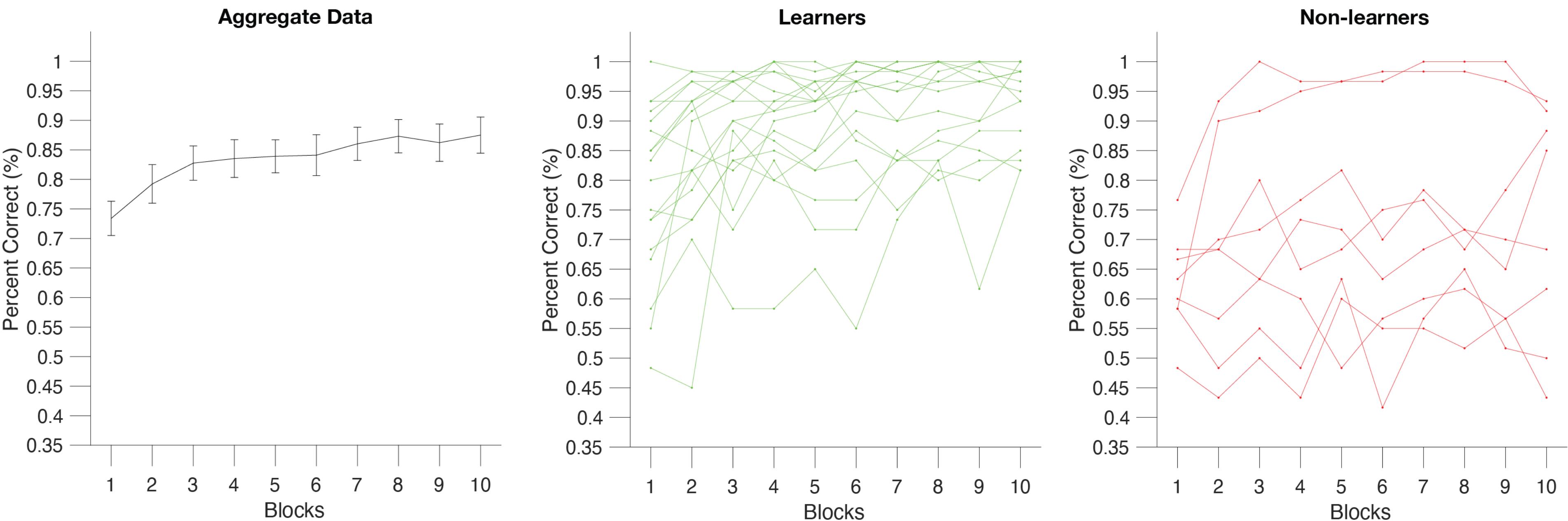


# Awareness Test

Only subjects that recreated all triplets were considered “aware”

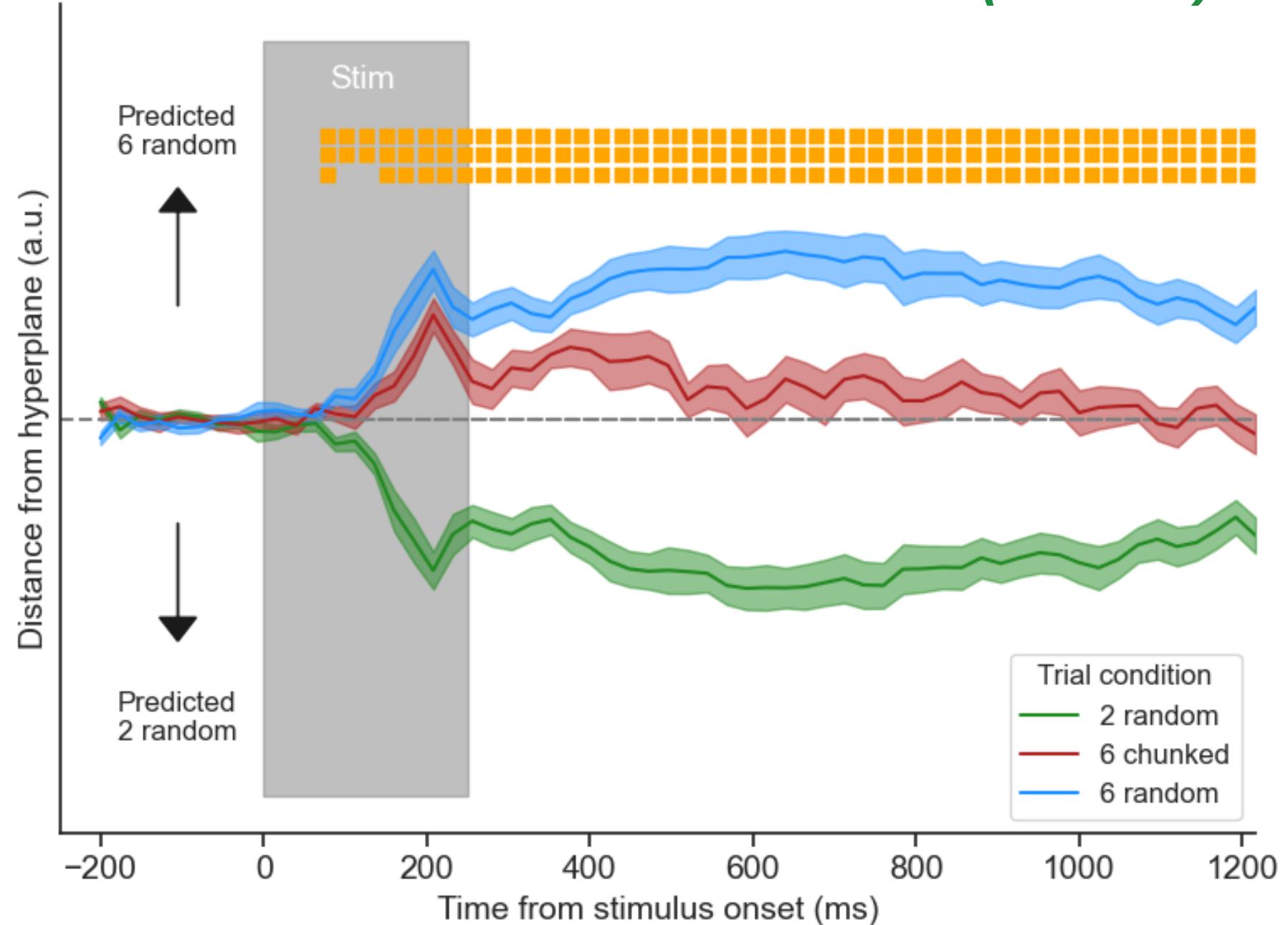


# Training Results

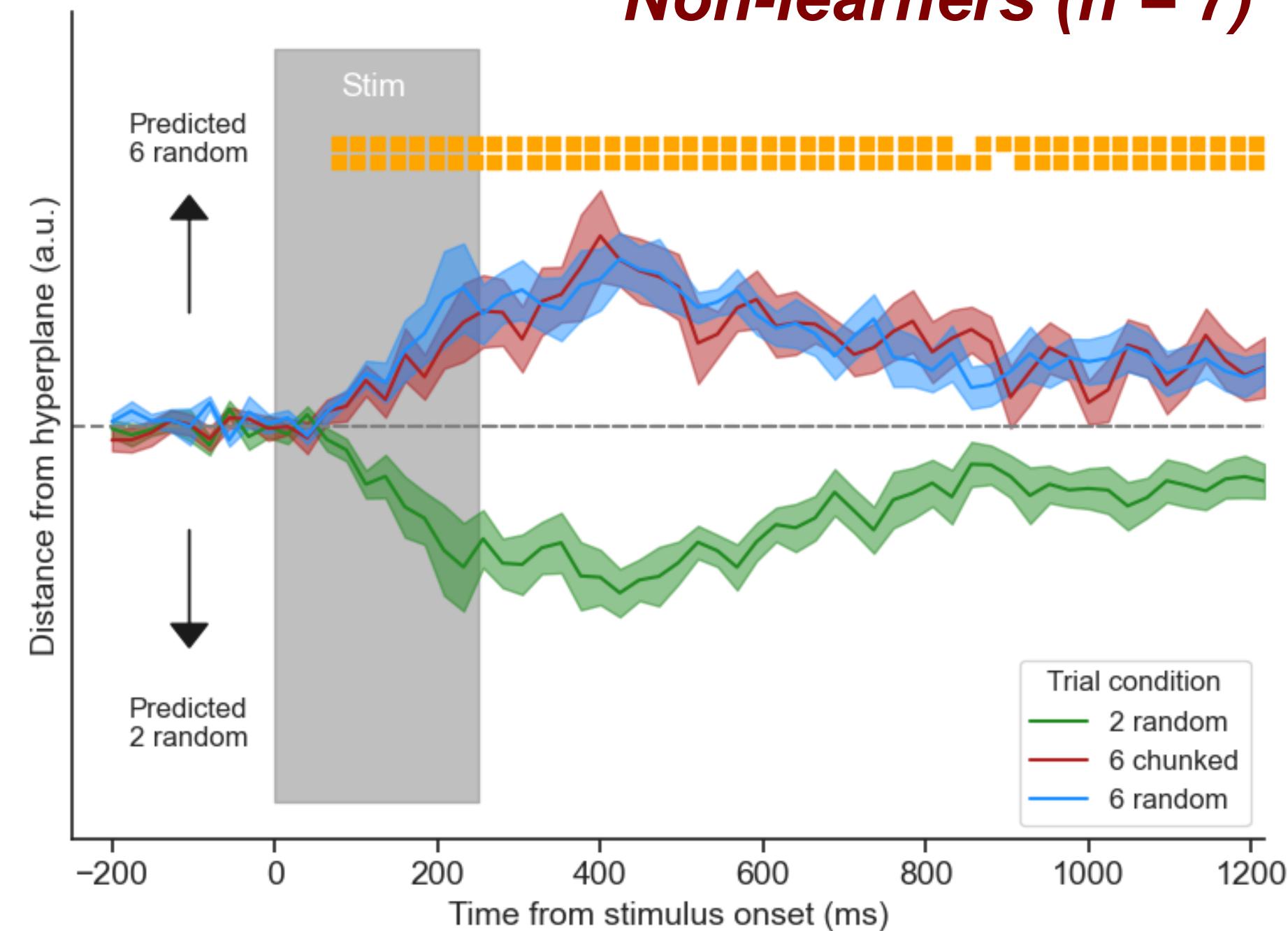


# Learners vs non-learners

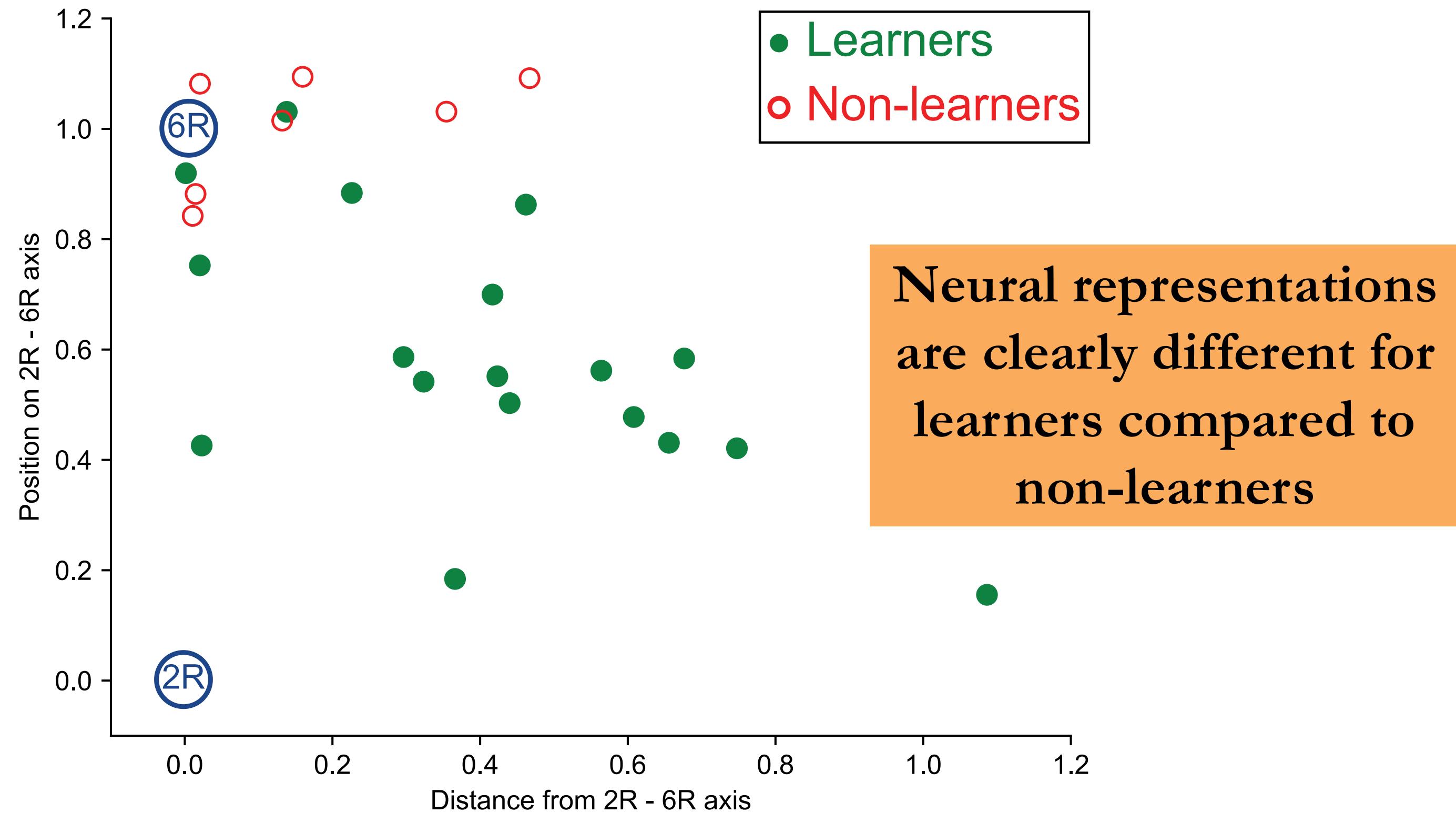
*Learners (n = 18)*



*Non-learners (n = 7)*

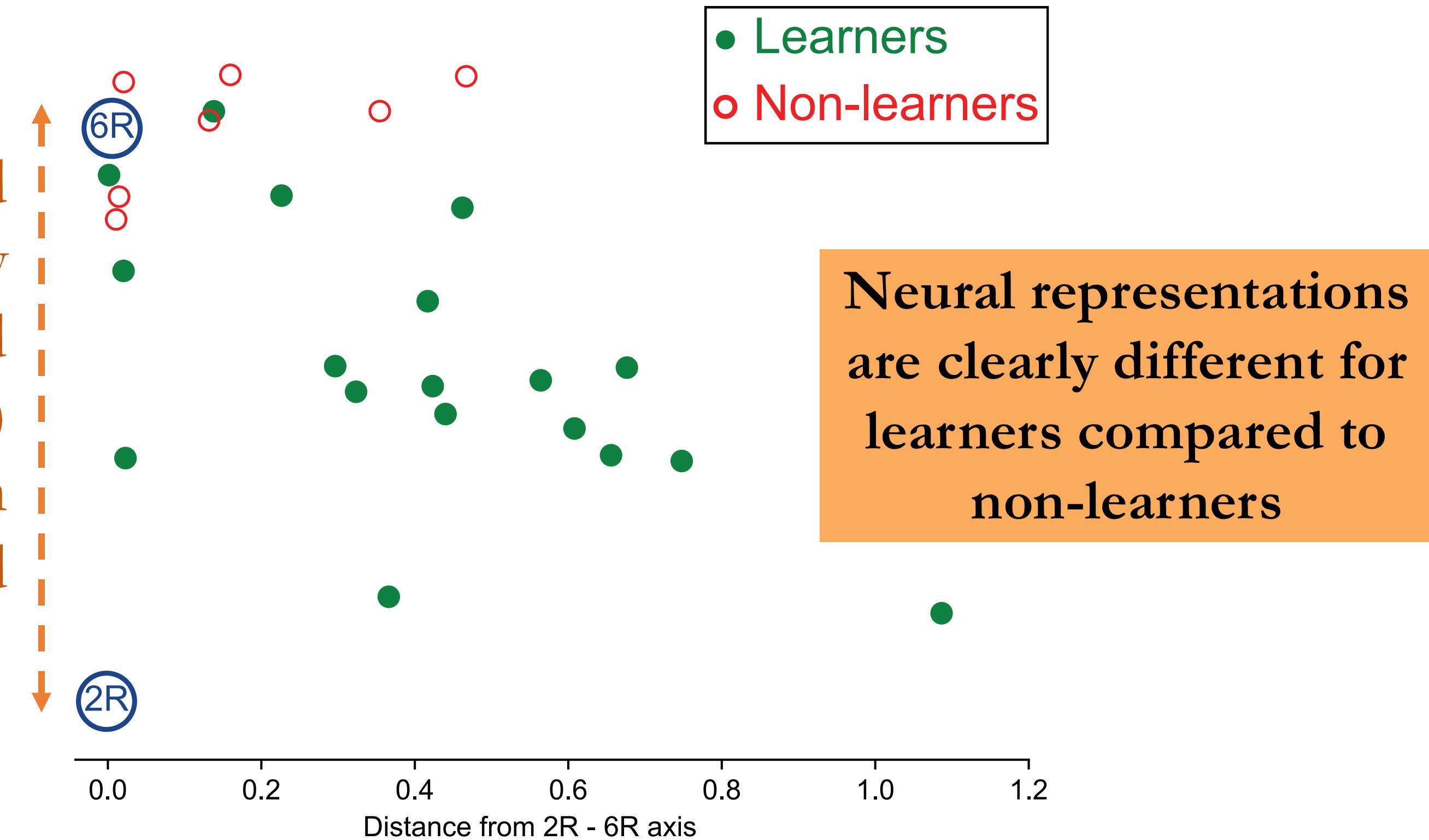


# Multidimensional scaling on each subject



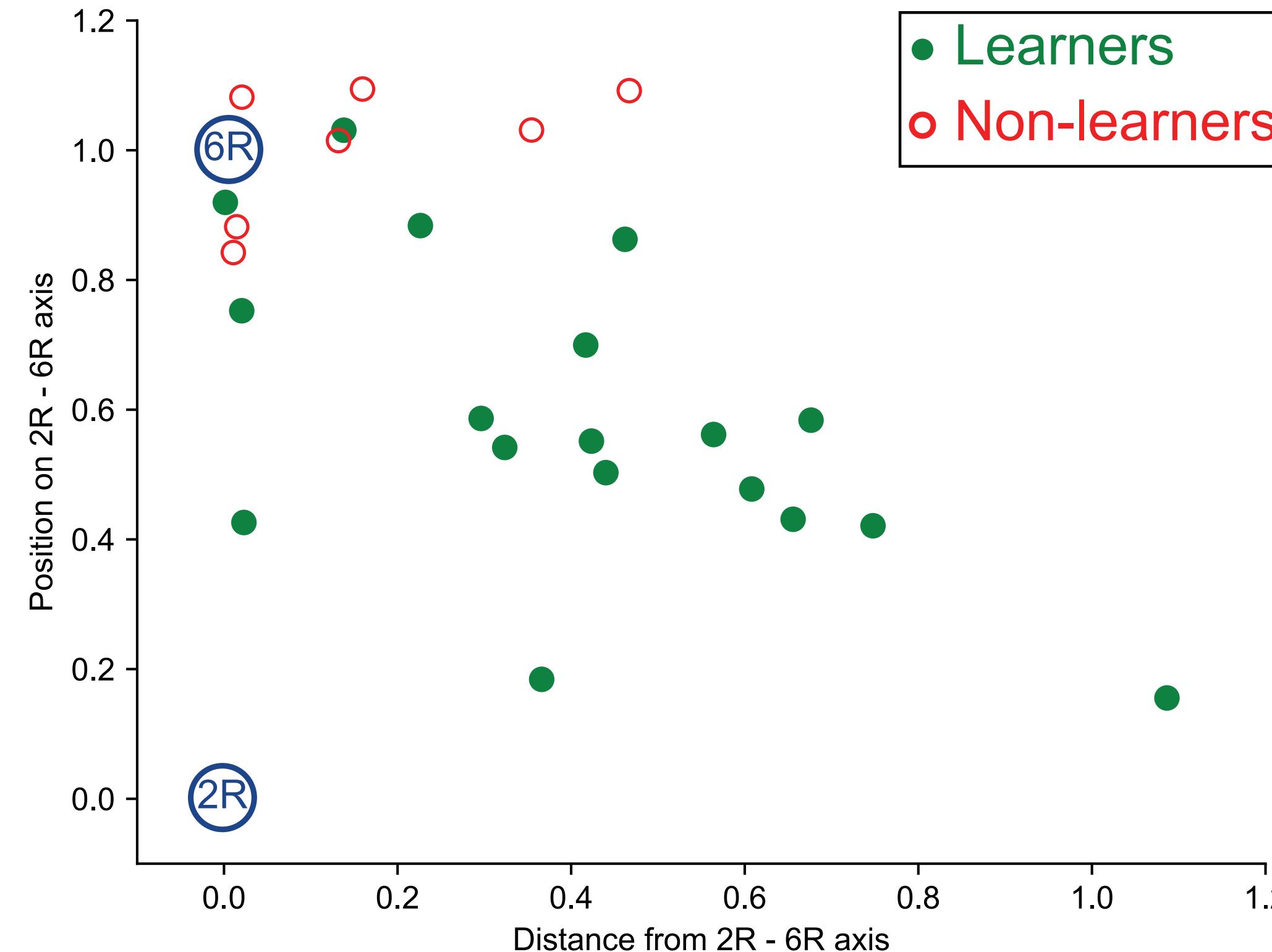
# Multidimensional scaling on each subject

We assume increased neural similarity towards 2R (and dissimilarity to 6R) suggests a reduction in working memory load



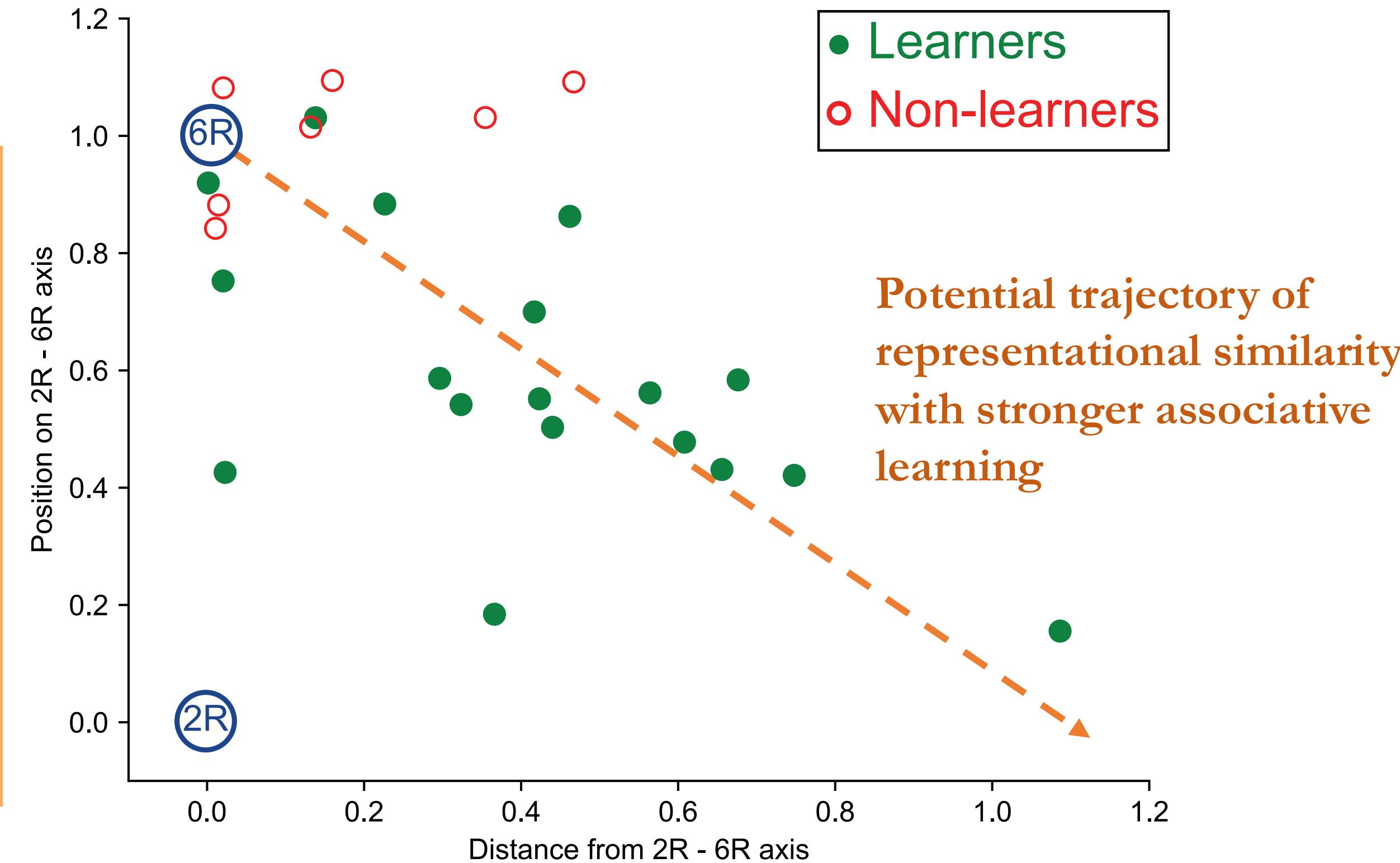
# Multidimensional scaling on each subject

We ought to have a cognitive model that **explains** the neural representation and **make specific predictions** of the representational similarity so that it is **falsifiable**.



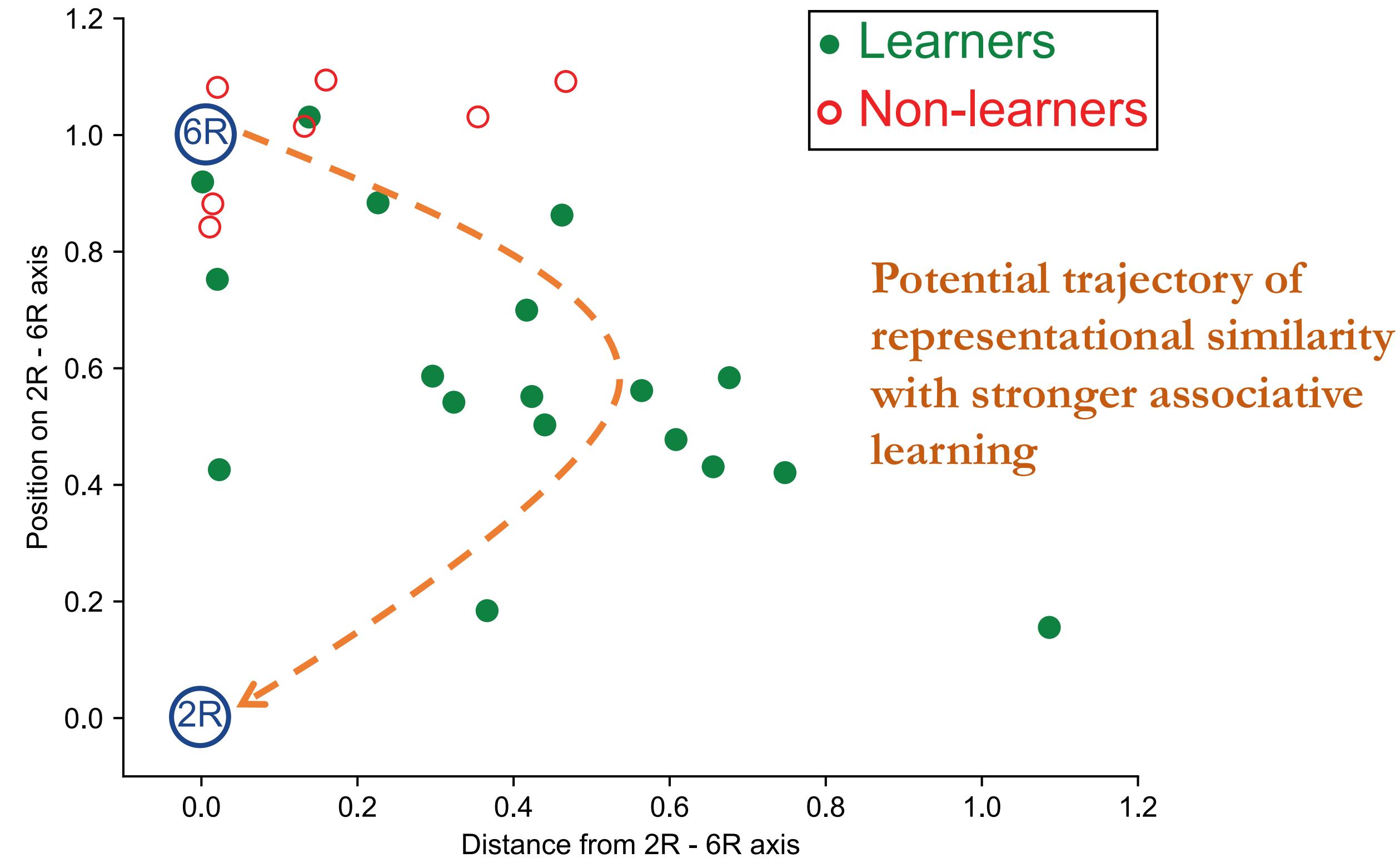
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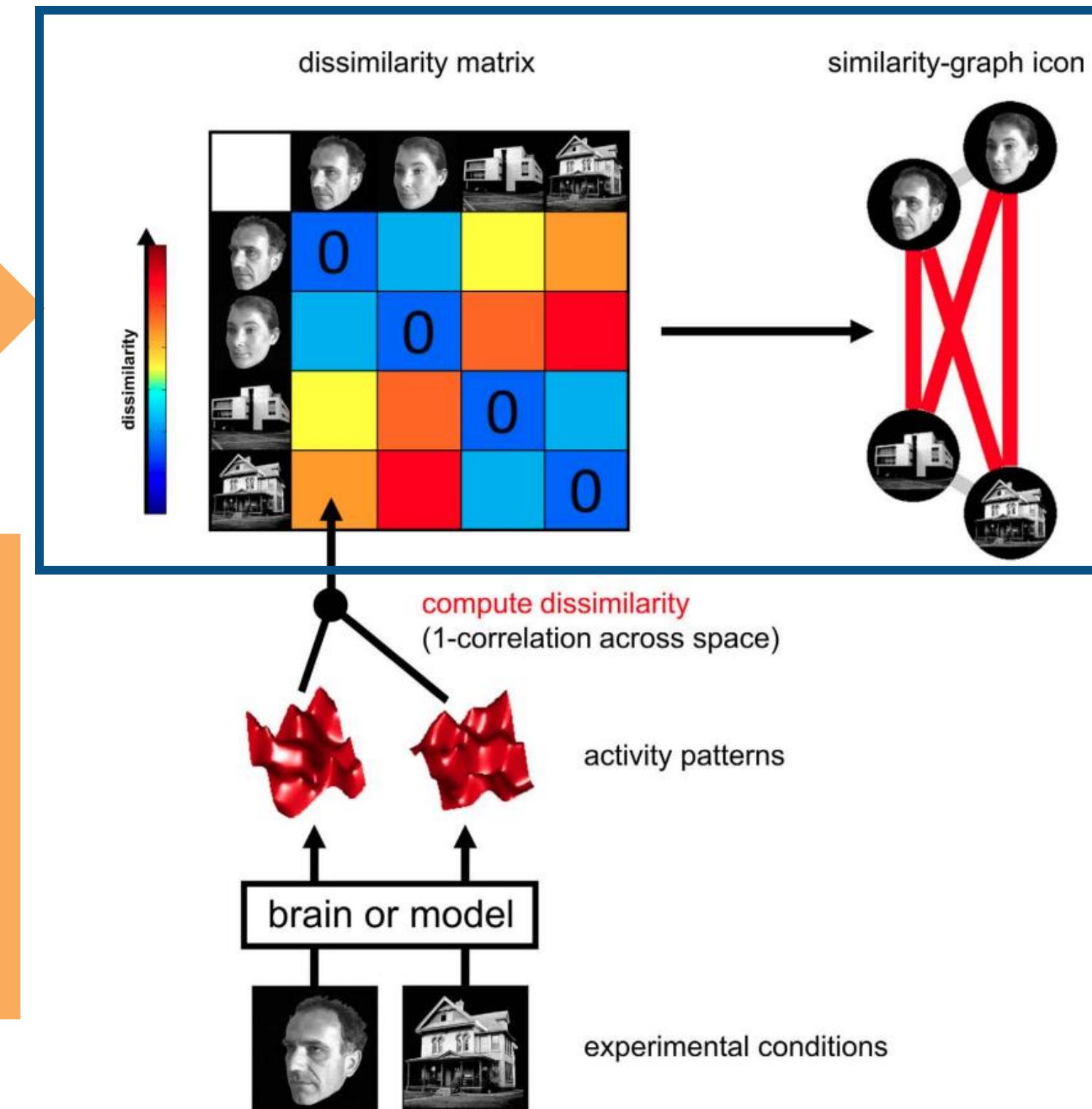
# Multidimensional scaling on each subject

We ought to have a cognitive model that **explains** the neural representation and **make specific predictions** of the representational similarity so that it is **falsifiable**.



# Representational similarity analysis

We need **cognitive models** to predict the observed representational (dis)similarity



**Do similarity ratings predict neural similarity?**

# Representations in formal cognitive models

Classic formal models of cognition would derive the psychological representation using multidimensional scaling (MDS) of **similarity judgments**

- Collect similarity ratings for pairs of items in the stimulus set
- Reduce those ratings into a representation that best preserves the distance between the items; the closer the items, the more similar.

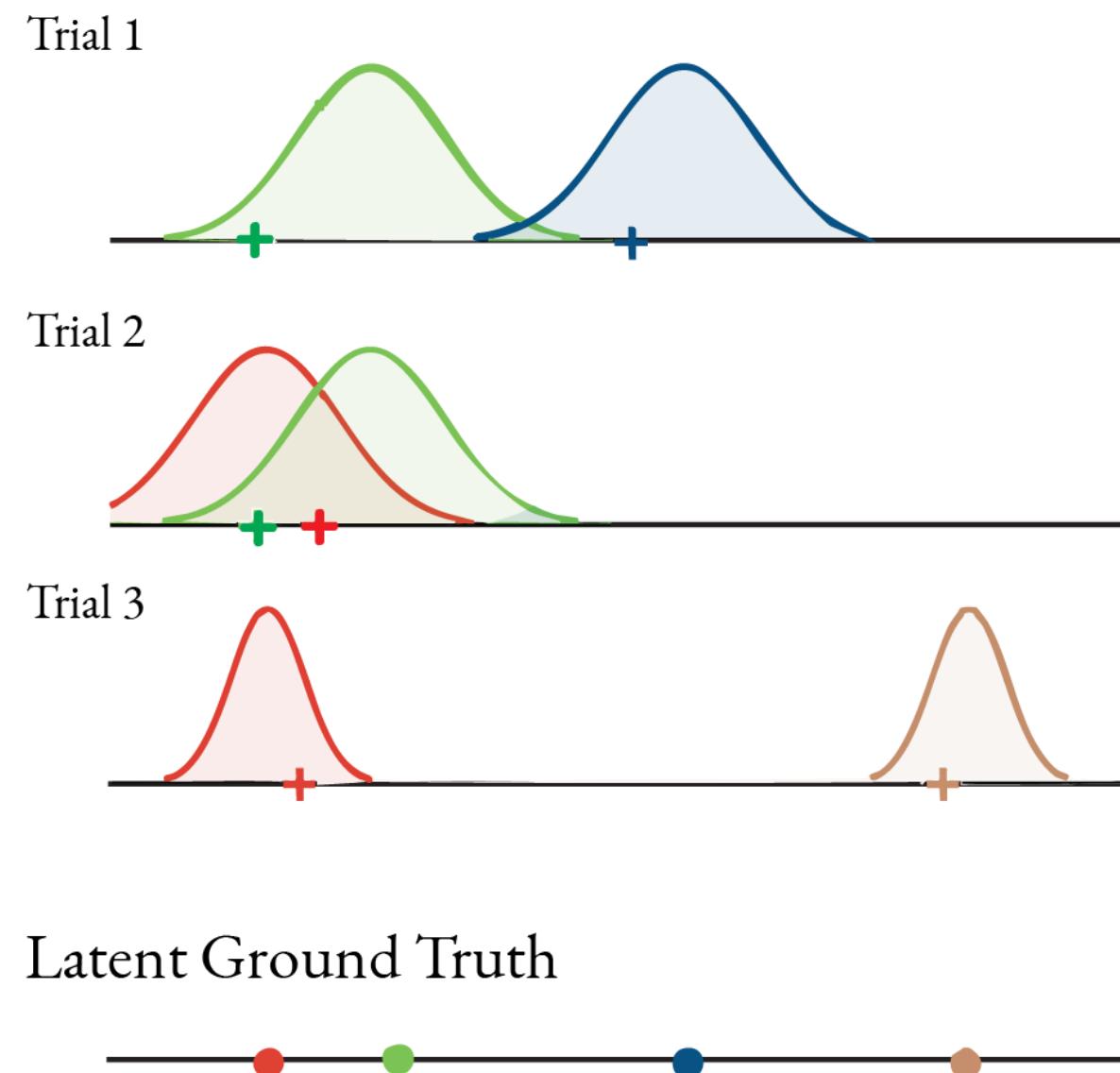
# Representations in formal cognitive models

For historical reasons, the similarity-based MDS-representation has been considered the psychological representation underlying cognition.

But **similarity judgments and cognition may not share the same mental representation.**

# Our modeling approach

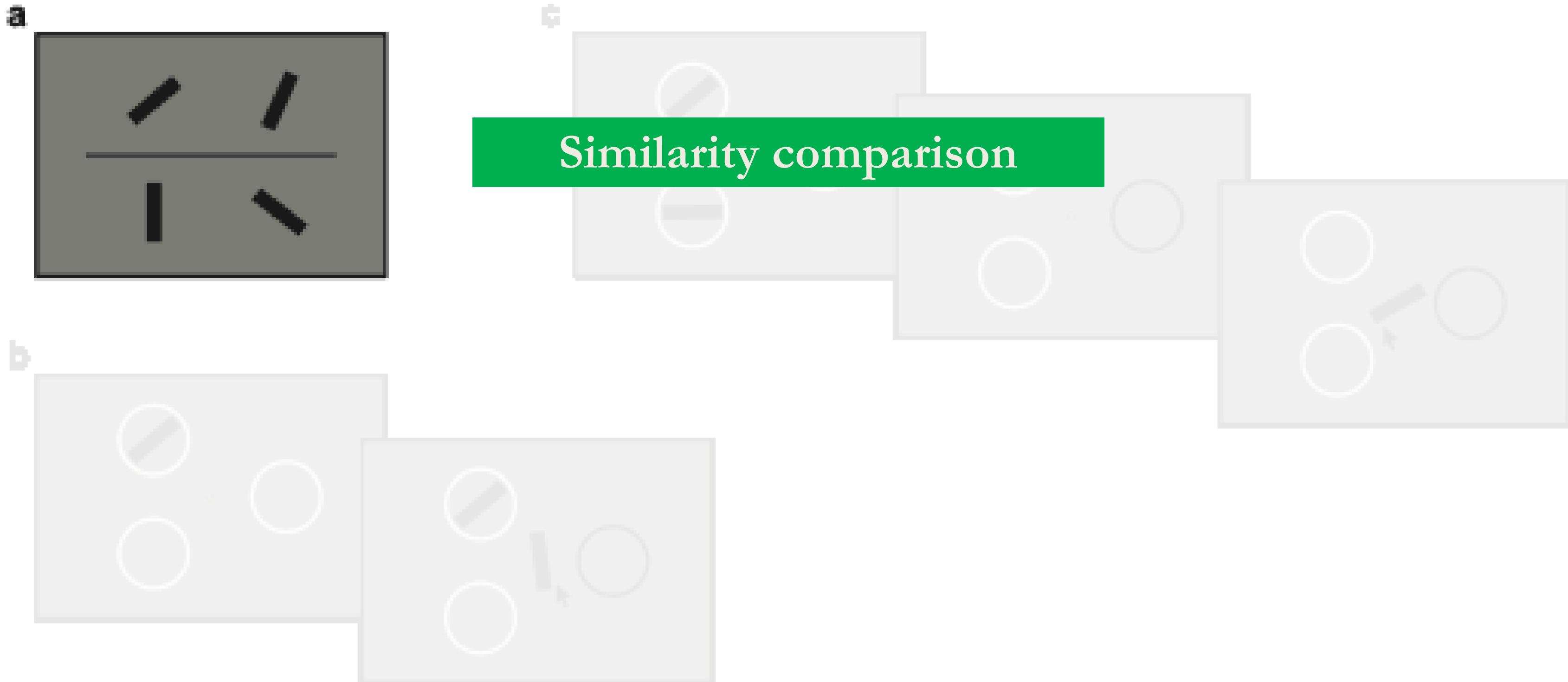
In brief, we used Bayesian MCMC methods to recover the **latent representation of oriented lines** used in three cognitive tasks.



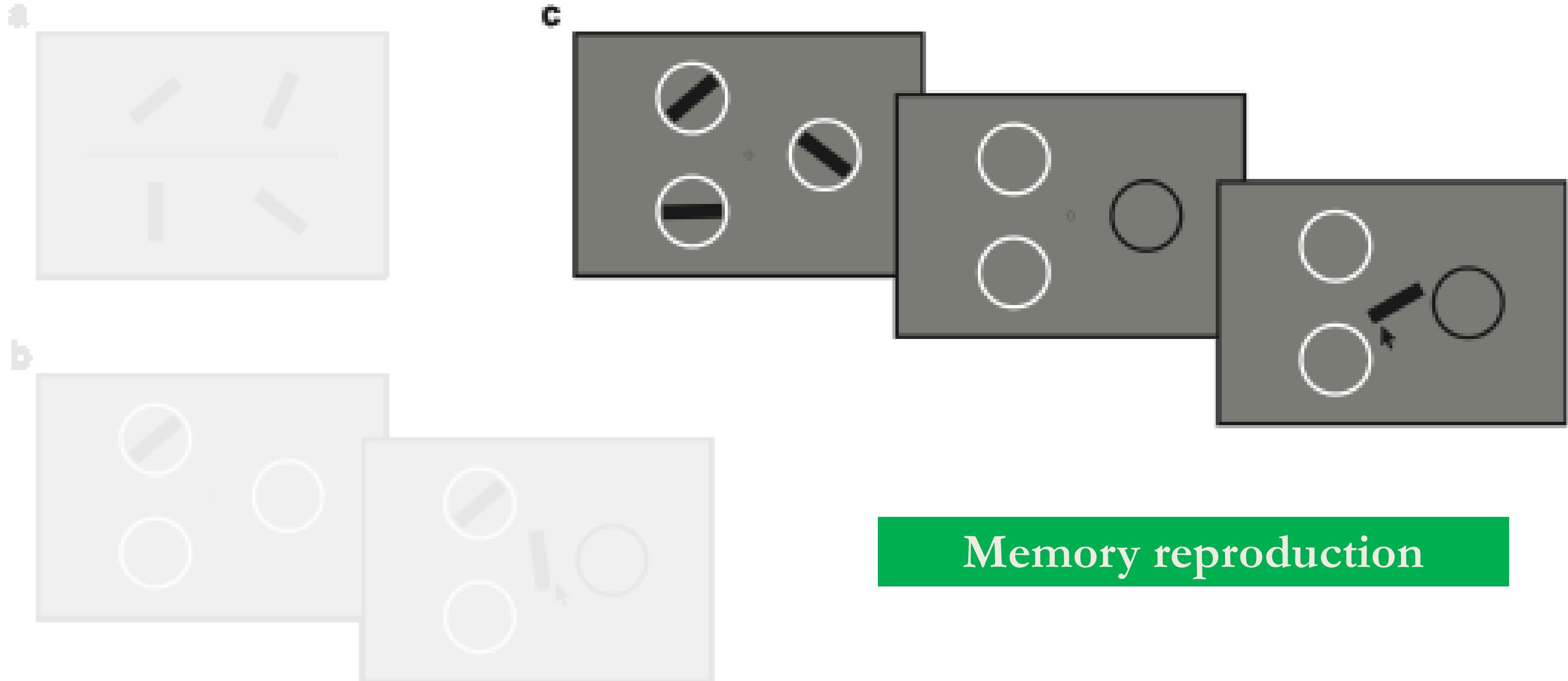
We use the observed similarity ratings to infer the latent representation of those items.



# The cognitive tasks



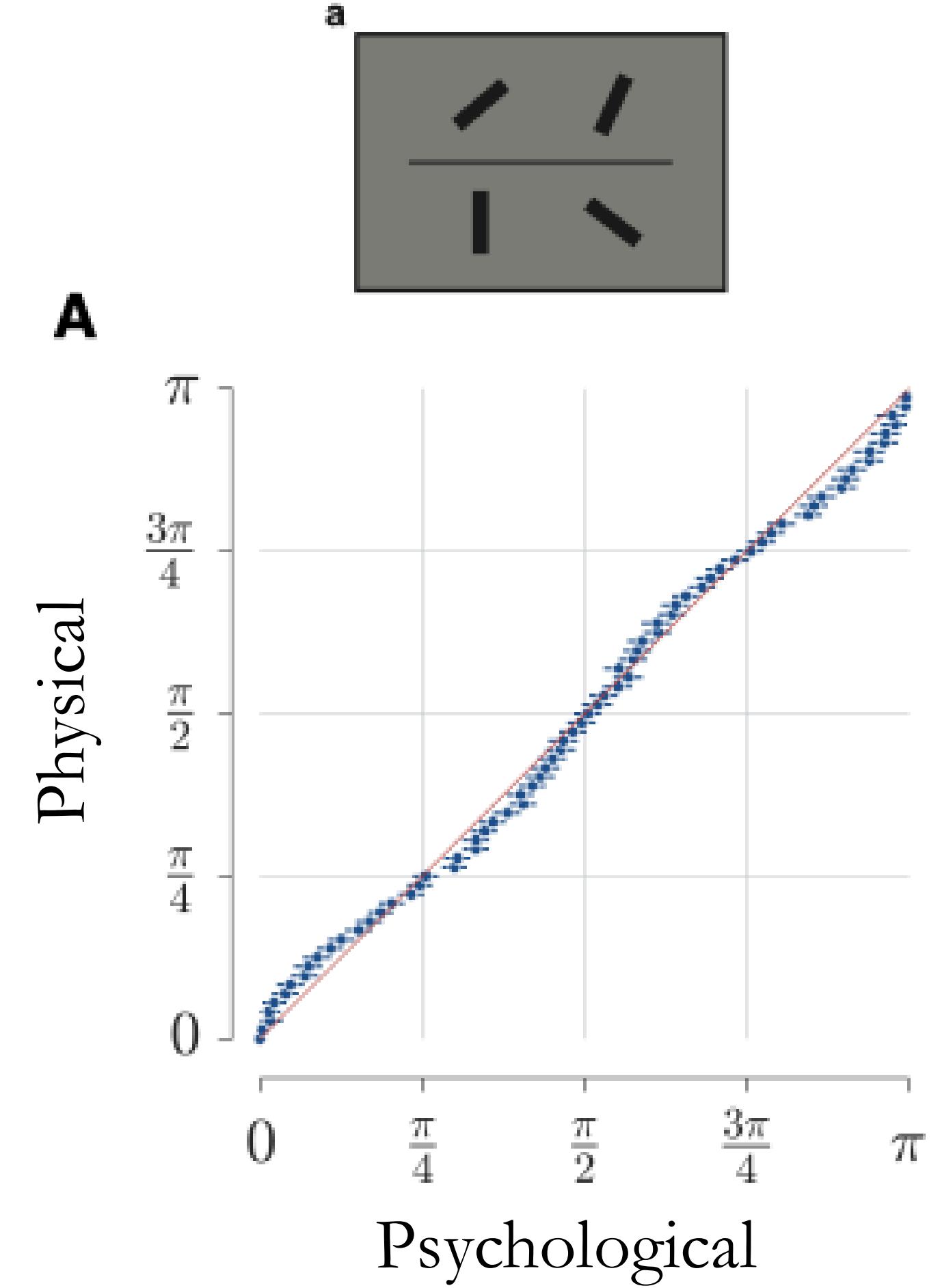
# The cognitive tasks



# Similarity comparison

The representation does not match the physical stimulus space – it is not exactly a diagonal line.

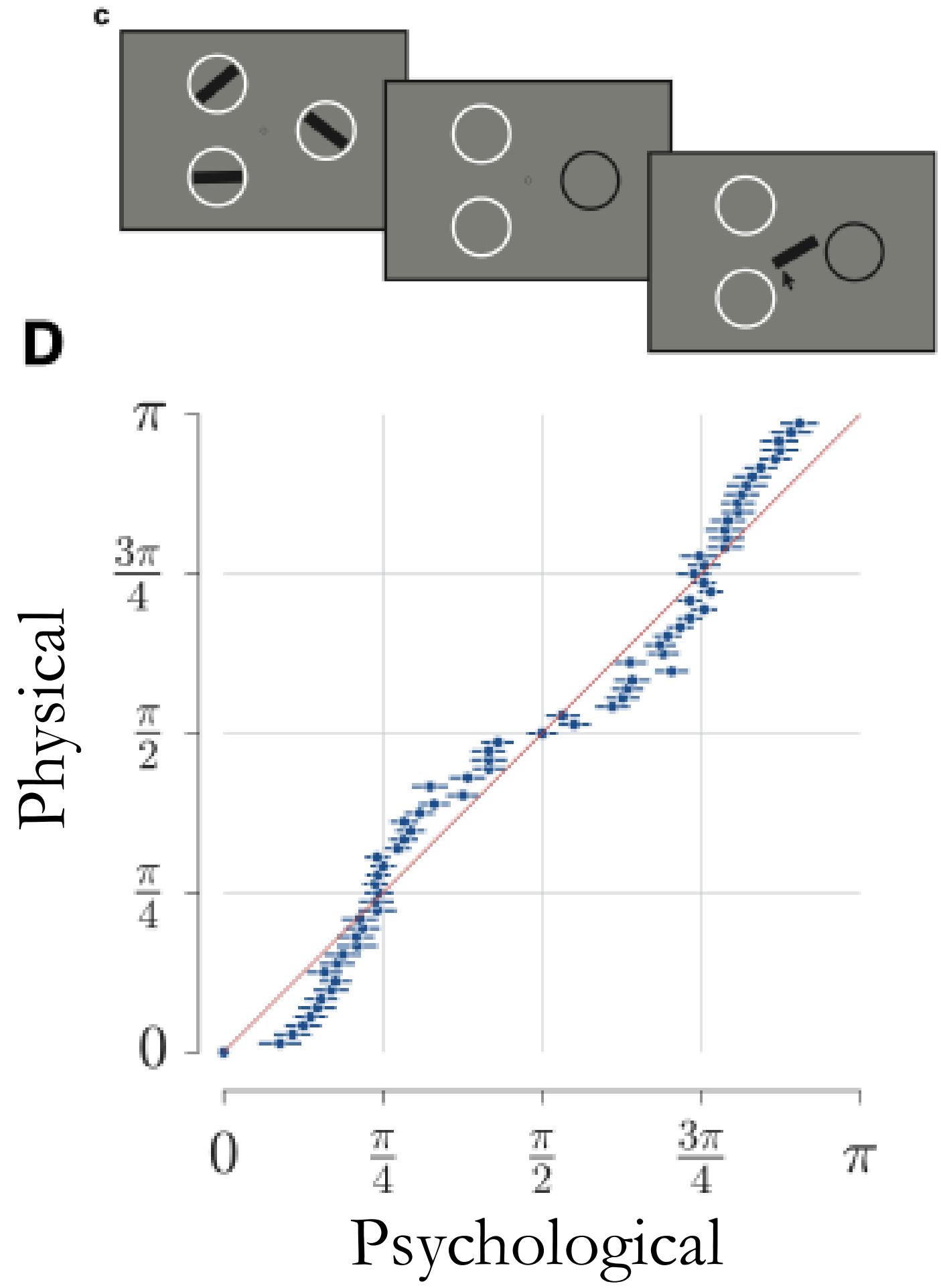
Clear deviations where close to vertical lines appear more vertical, and close to horizontal lines appear more horizontal.



# Memory reproduction

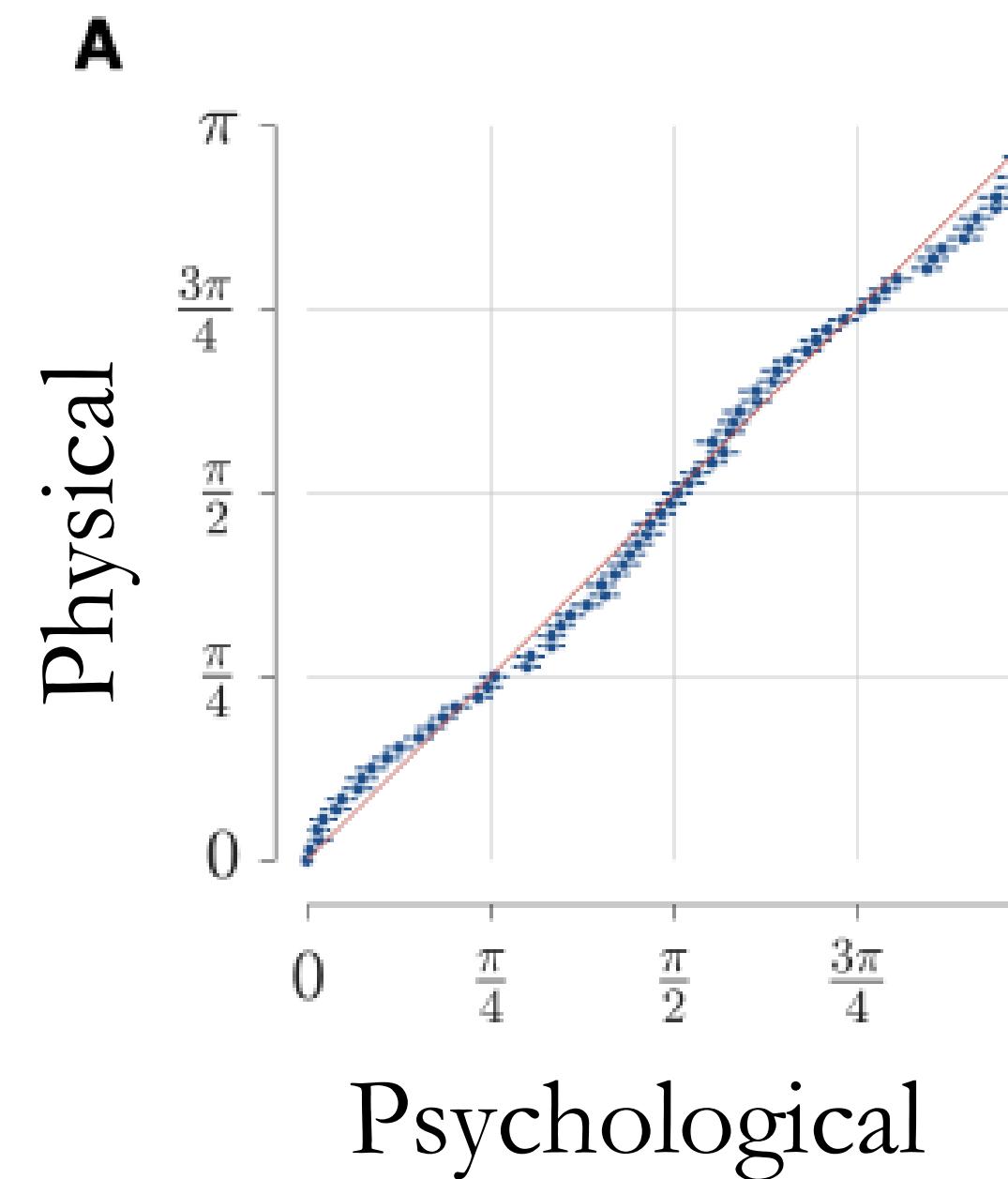
The representation for oriented lines do not exactly match the physically presented stimulus.

In working memory, the oriented lines are represented more towards the oblique directions than they actually are.

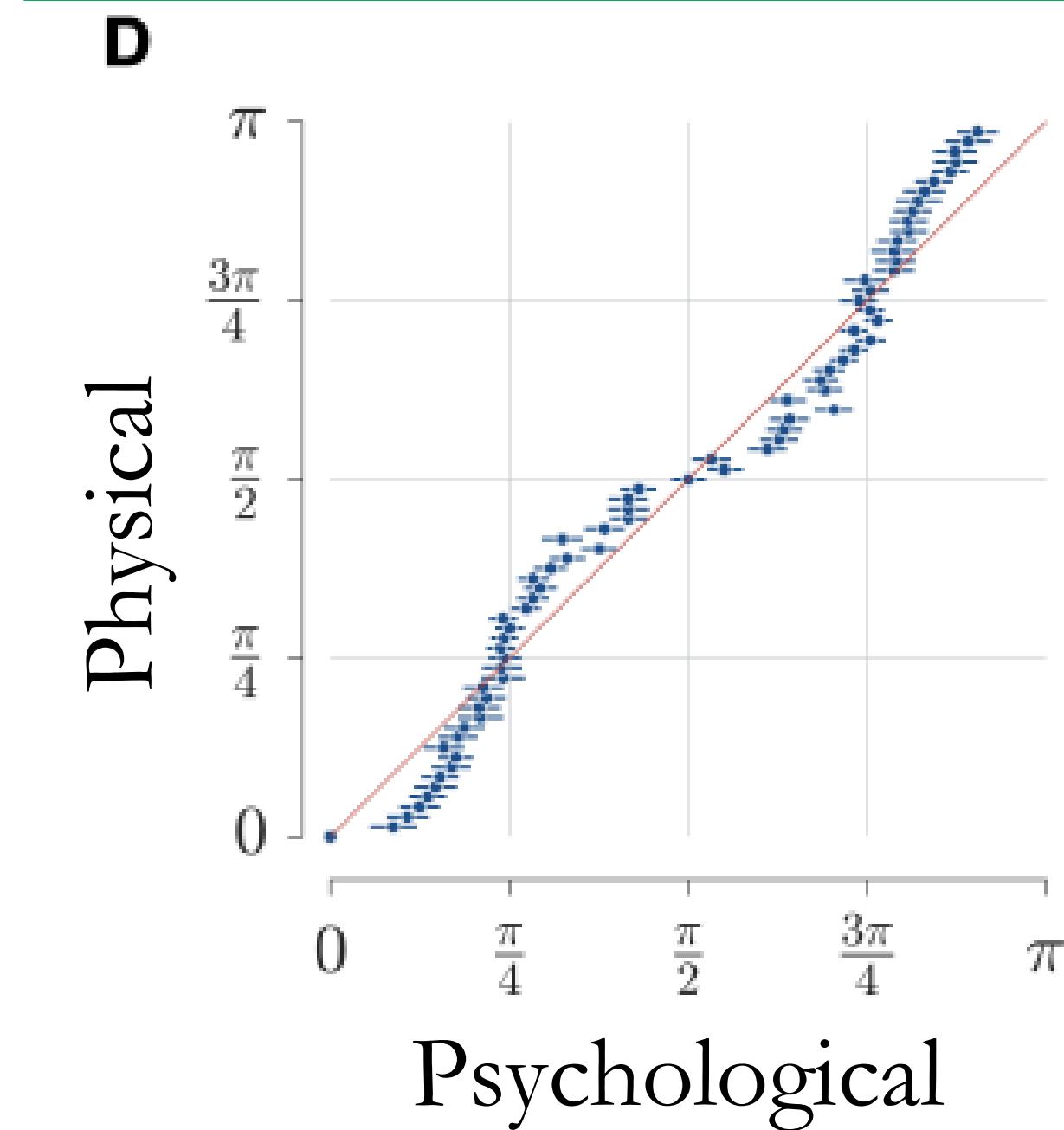


Similarity comparisons and both reproduction tasks **do not**  
share the same cognitive representation.

Similarity comparison



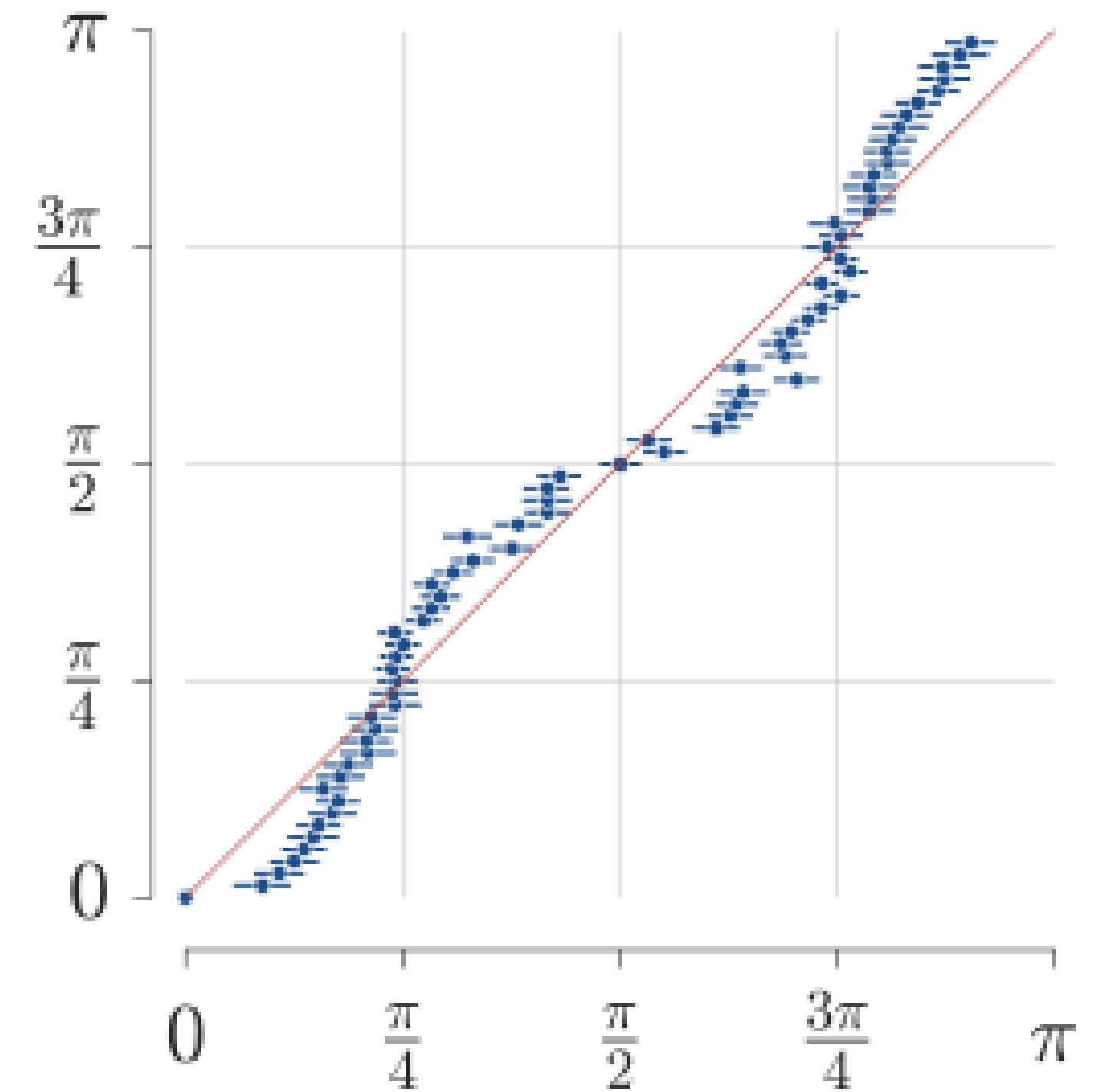
Memory reproduction



# Does psychological similarity predict neural similarity?

Psychological similarity cannot be assumed to be the basis for cognition (and **possibly not the basis for similarity of neural representations**).

D



# Takeaways

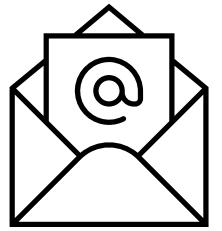
We ought to have formal cognitive models that predict and explain the similarity of neural representations.

We cannot assume that psychological similarity will predict neural similarity as the cognitive representation may not be the same.

How do we integrate formal models of cognition with empirical neuroscience? Come to **MathPsych** next year to find out...

# Thank you!

Preprint



[wiliam.ngiam@adelaide.edu.au](mailto:wiliam.ngiam@adelaide.edu.au)



<https://palm-lab.github.io>



[@williamngiam.github.io](https://@williamngiam.github.io)

