

Multivariate classification shows associative learning reduces working memory load

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Symposium: A model collaboration

AUSTRALASIAN COGNITIVE NEUROSCIENCE SOCIETY

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A quick introduction to me

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

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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

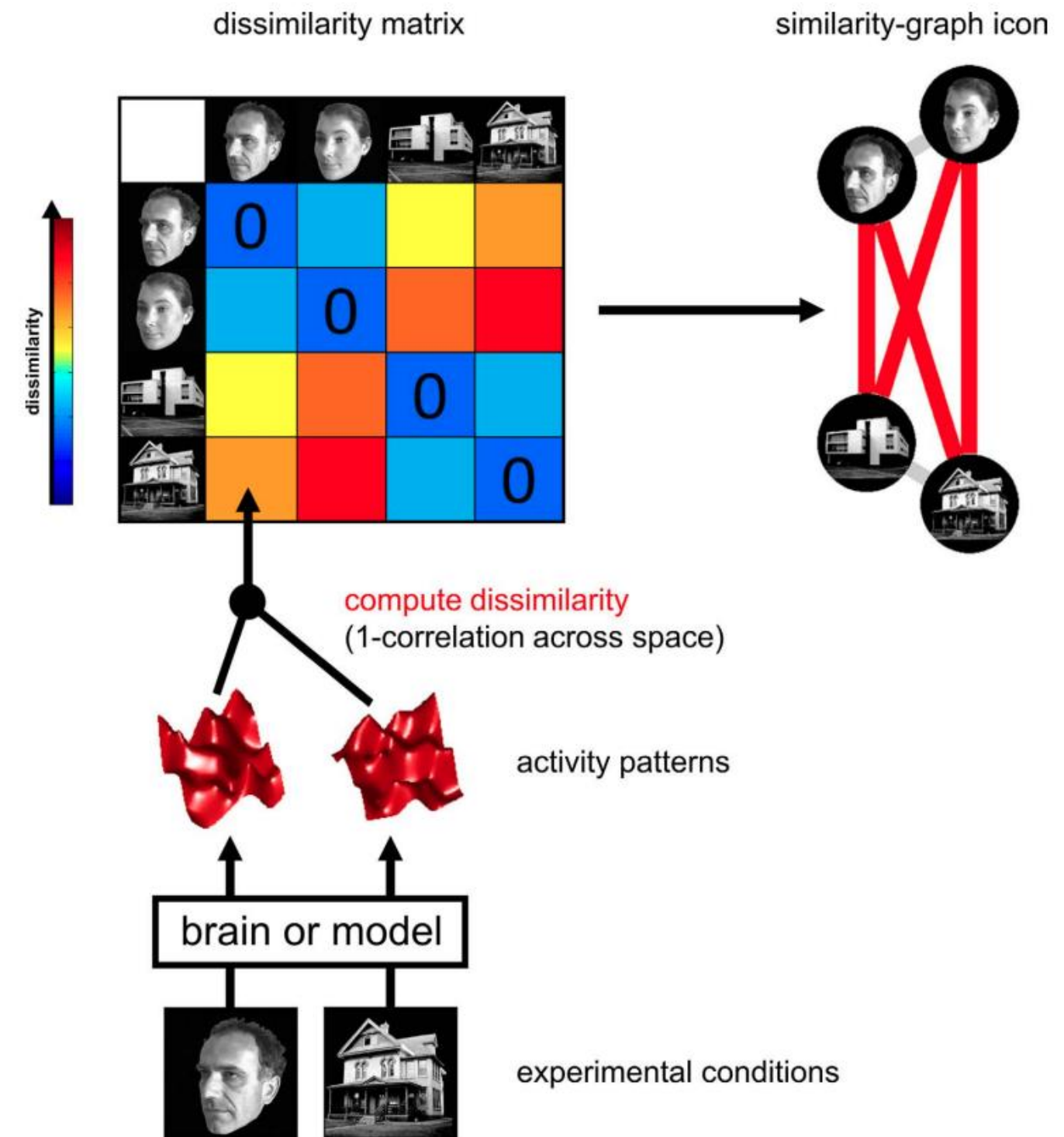
Cognitive theories ought to make predictions about neural signals and decoding results, such as those from **representational similarity analysis**.

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

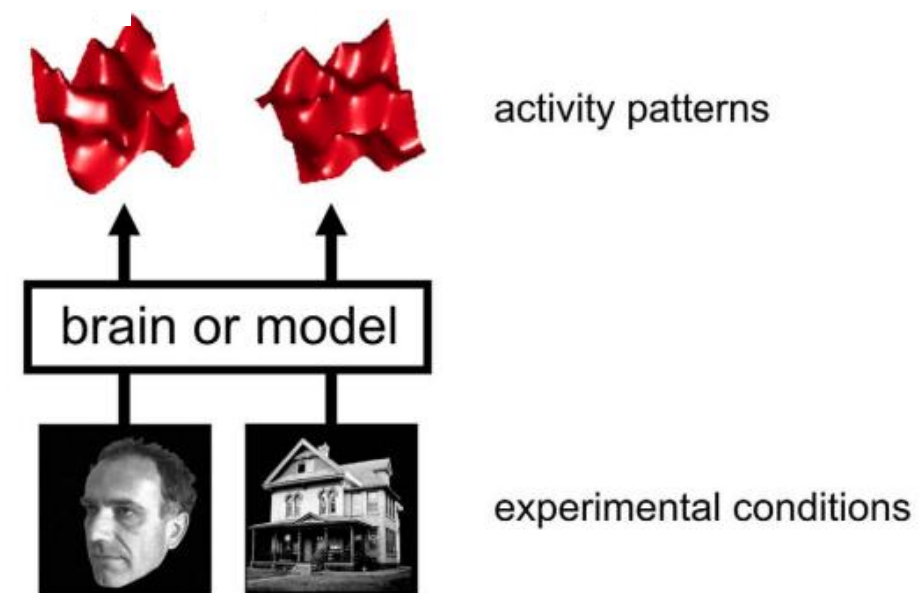
Other neuroscientists are using training neural network models to achieve human-like performance or using machine learning to decode neural representations.

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



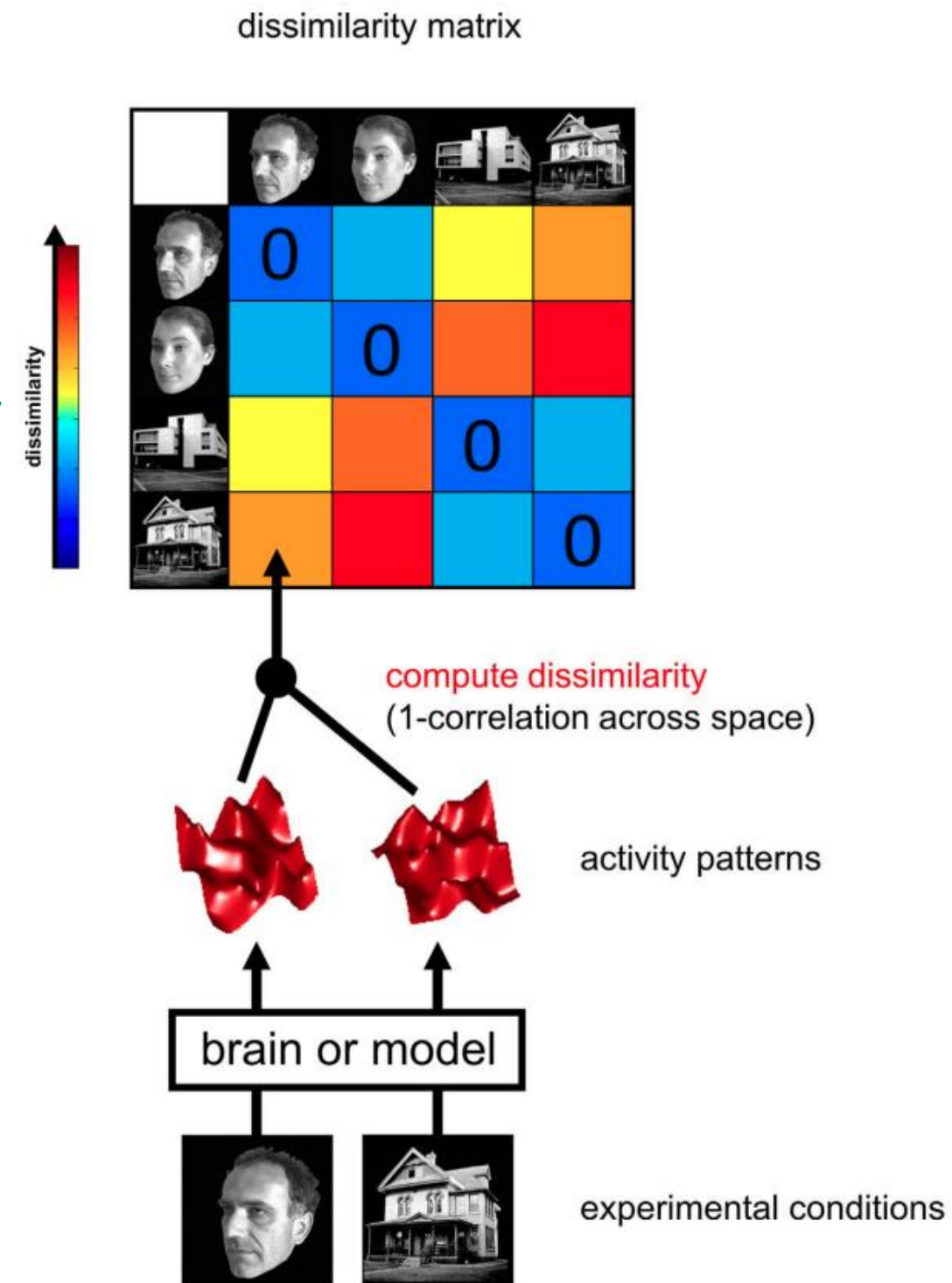
Representational similarity analysis

The researcher collects
neuroimaging data under
experimental conditions

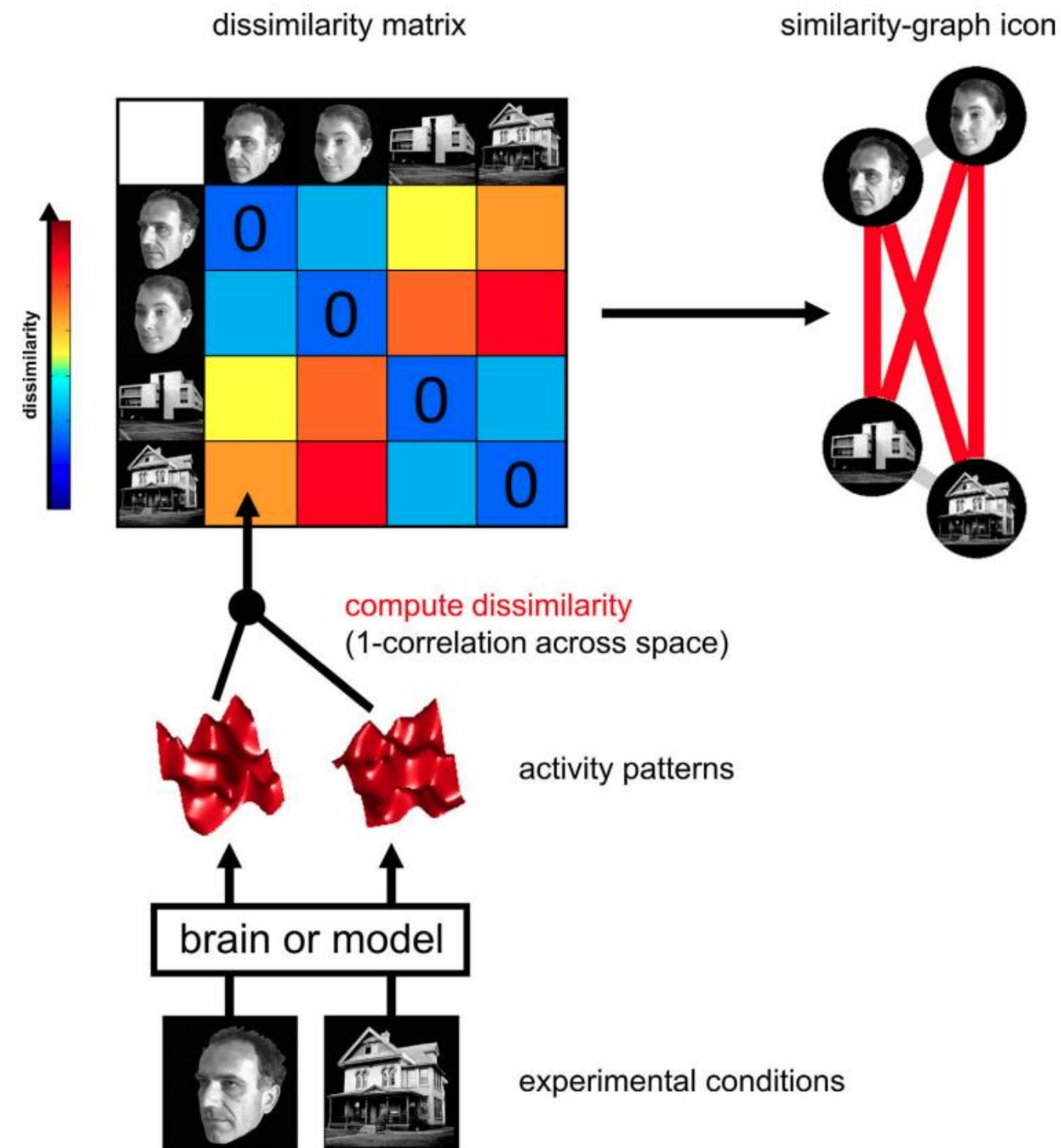


Representational similarity analysis

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

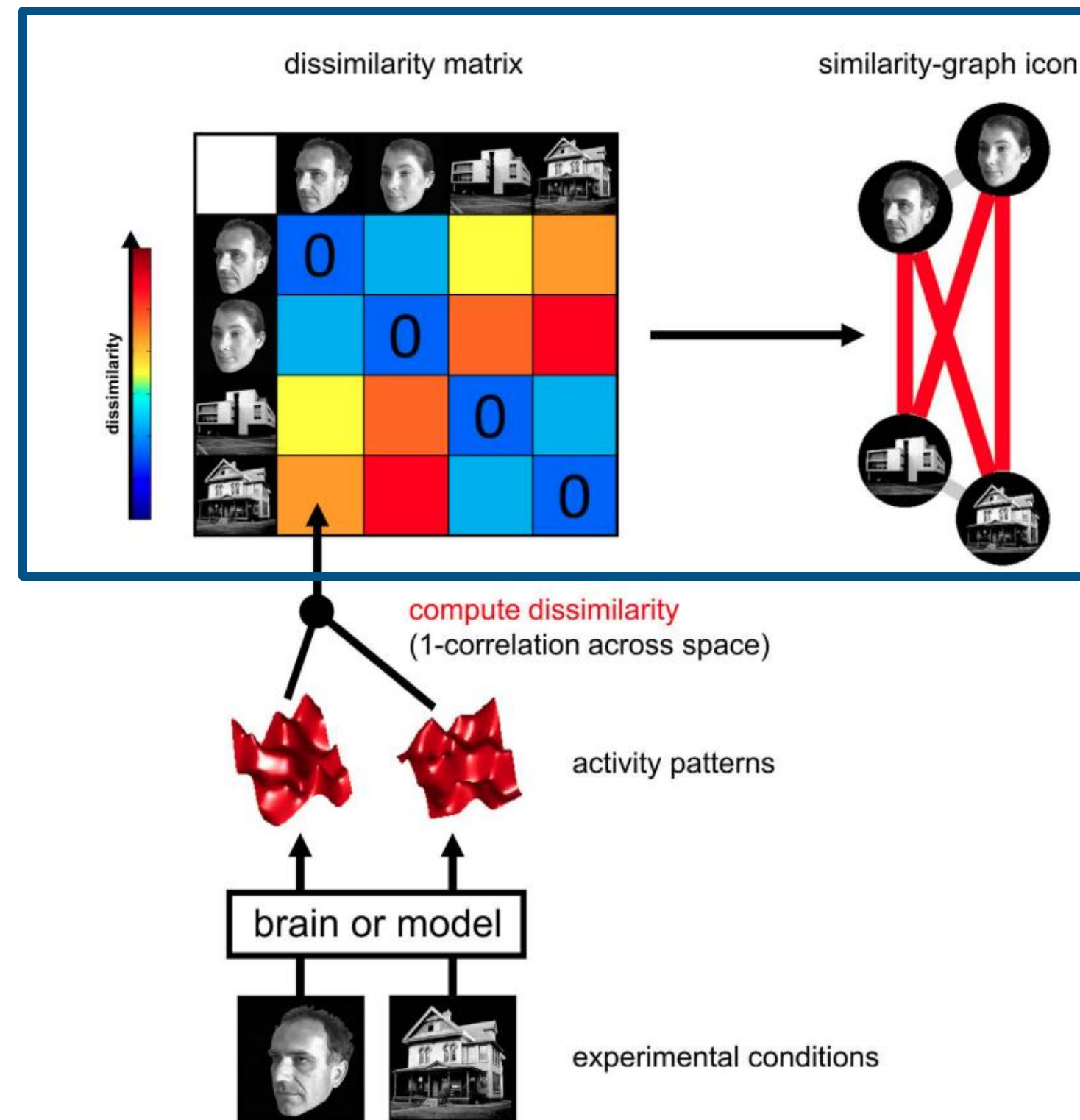


Representational similarity analysis



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

Representational similarity analysis

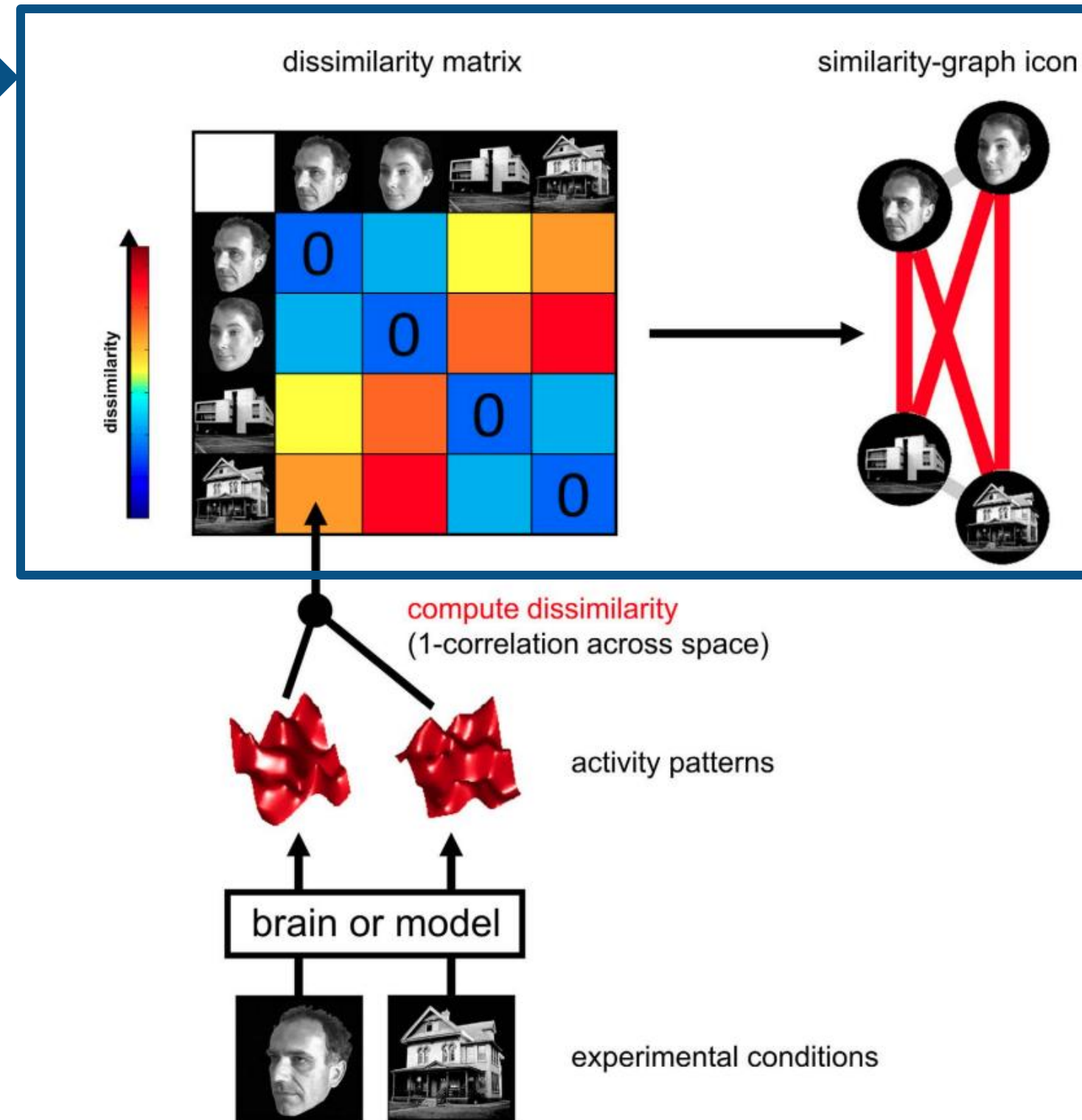
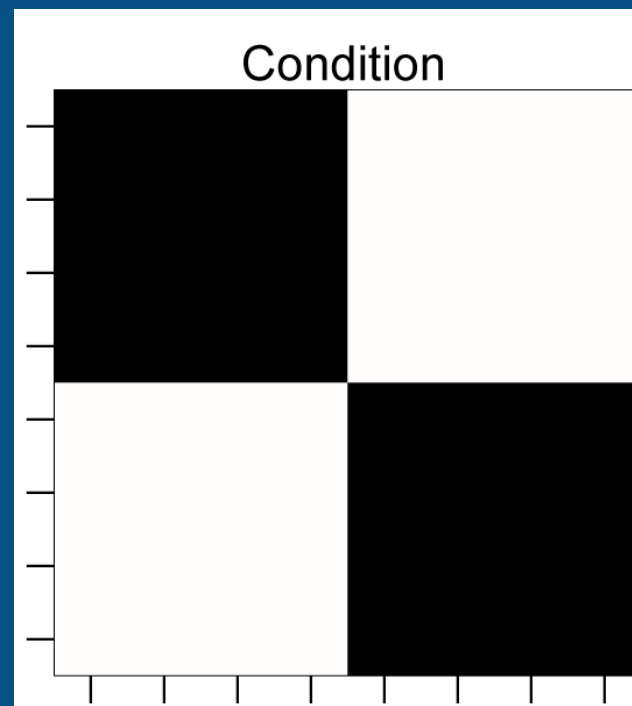


Do these results provide
a veridical basis for
models of cognition?

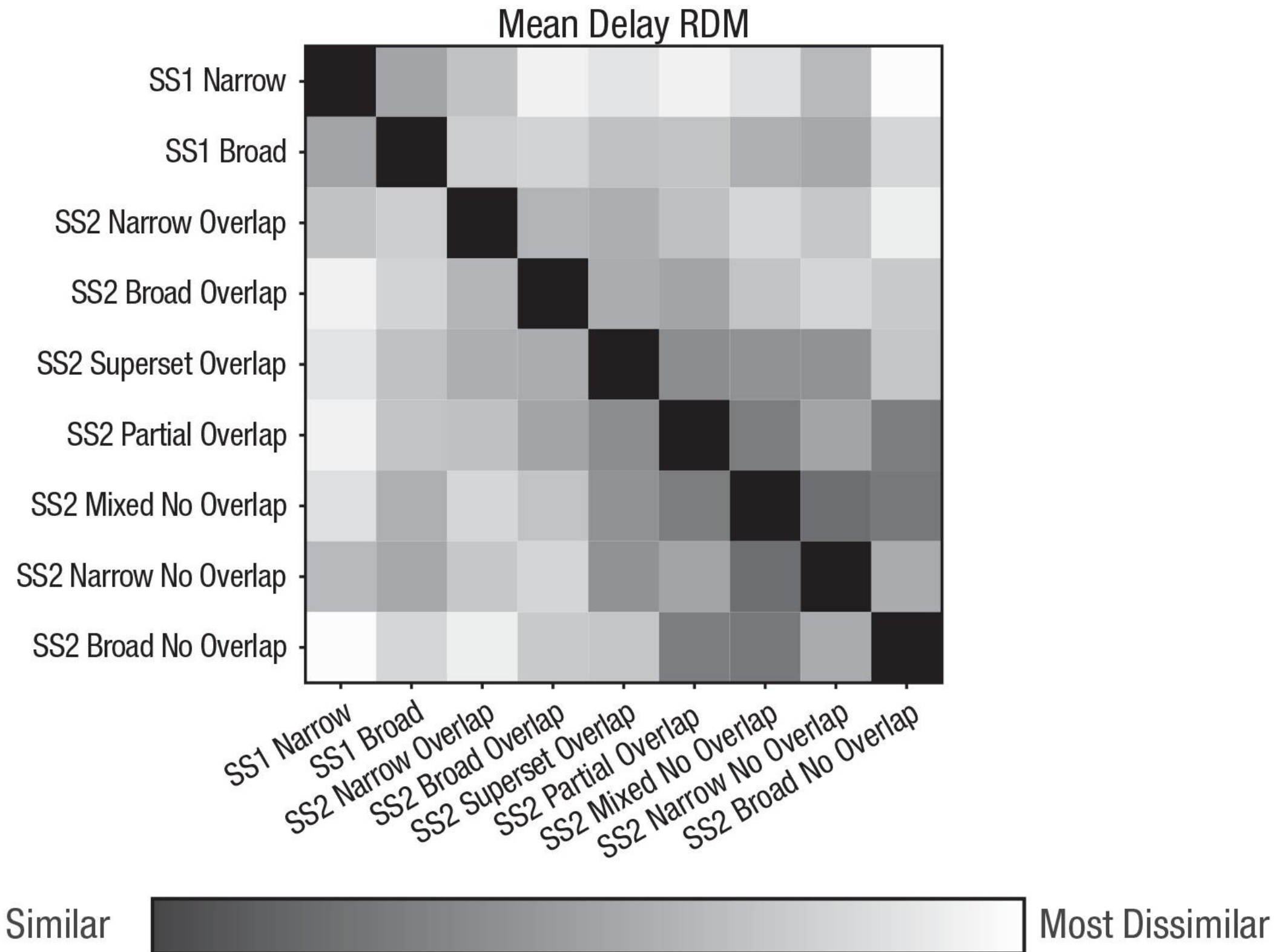
We might be **HARKing**
and/or drawing ad-hoc
conclusions.

Representational similarity analysis

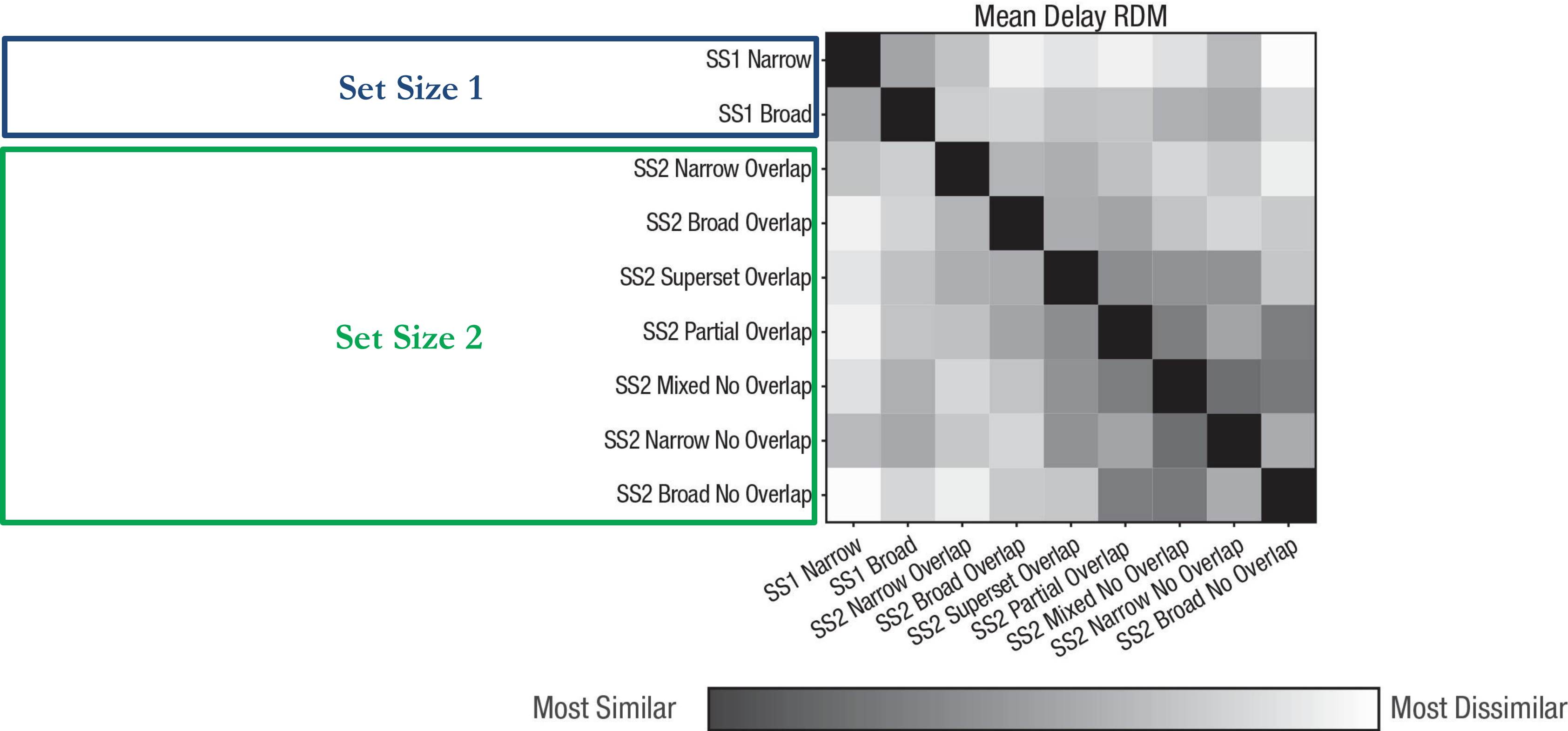
We might create models to predict the observed representational (dis)similarity



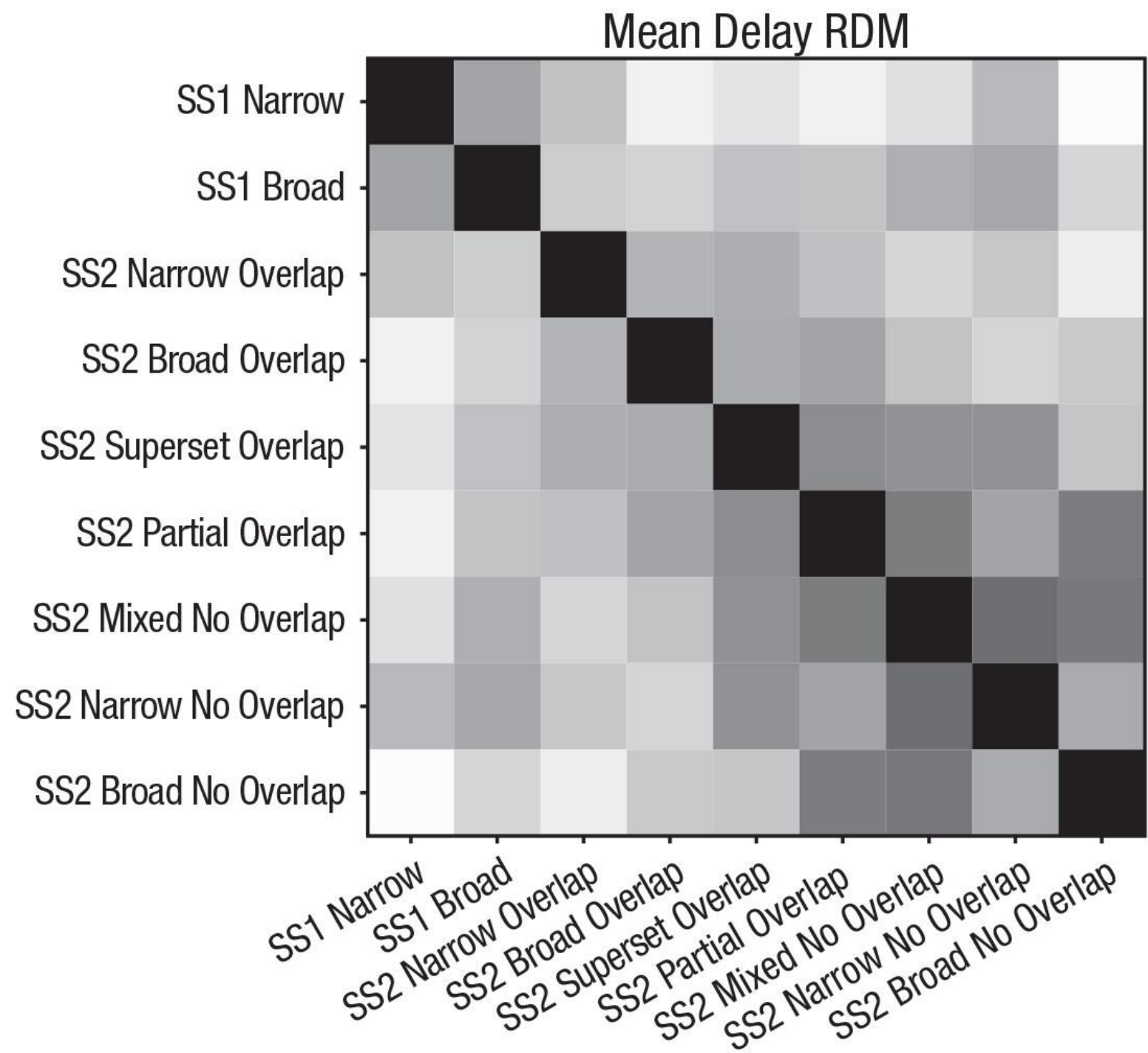
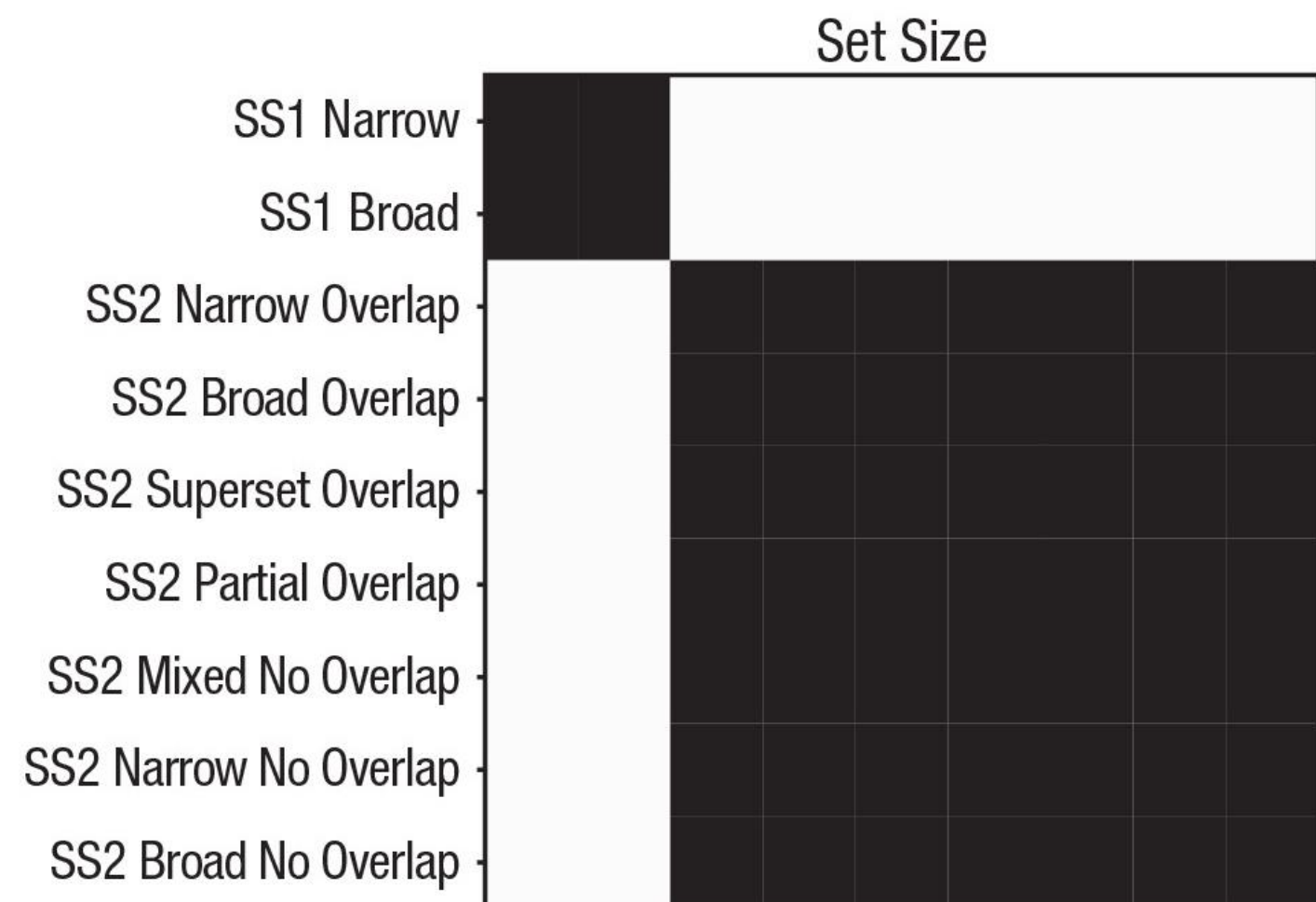
Decoding load in working memory



Decoding load in working memory



Decoding load in working memory

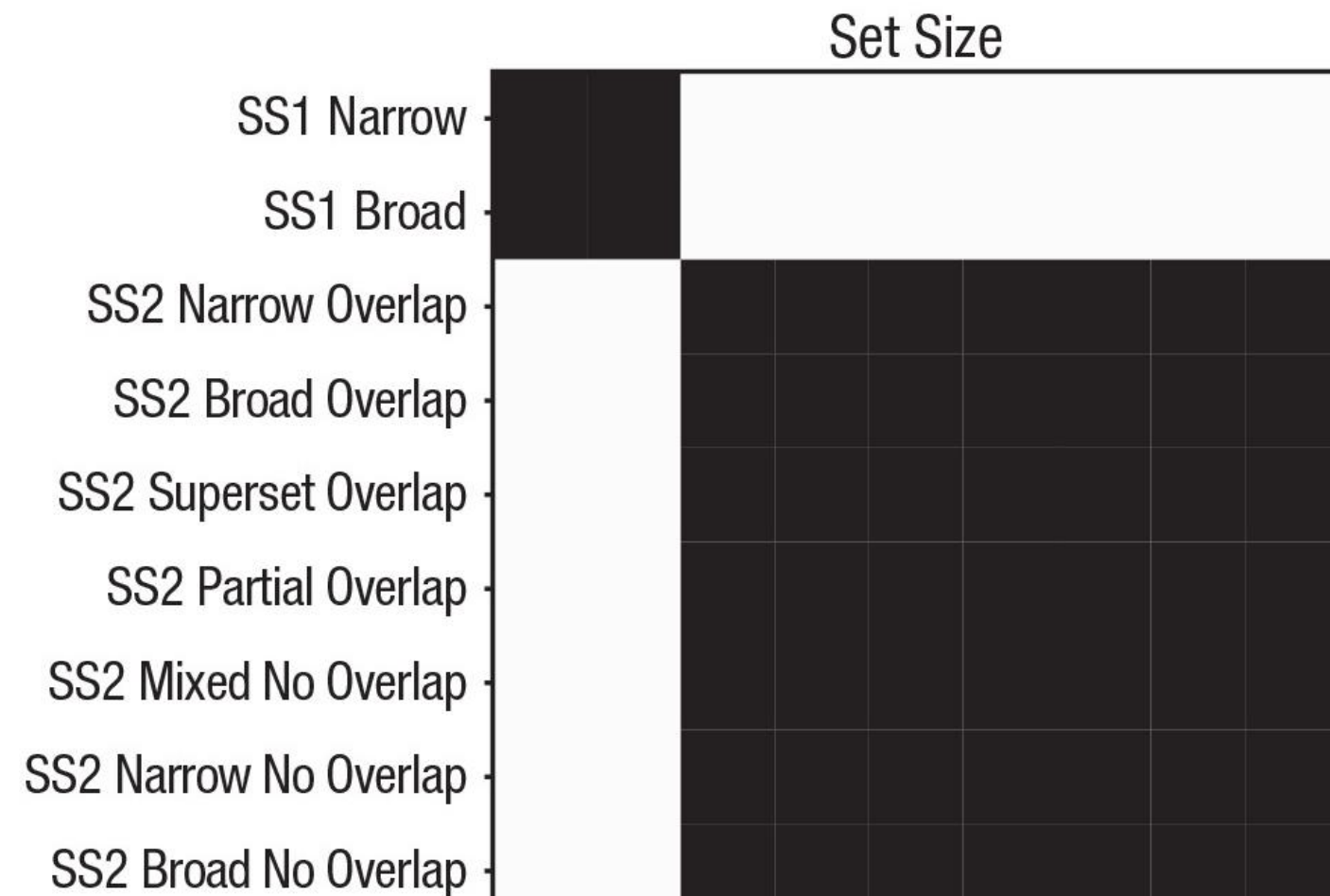


Most Similar

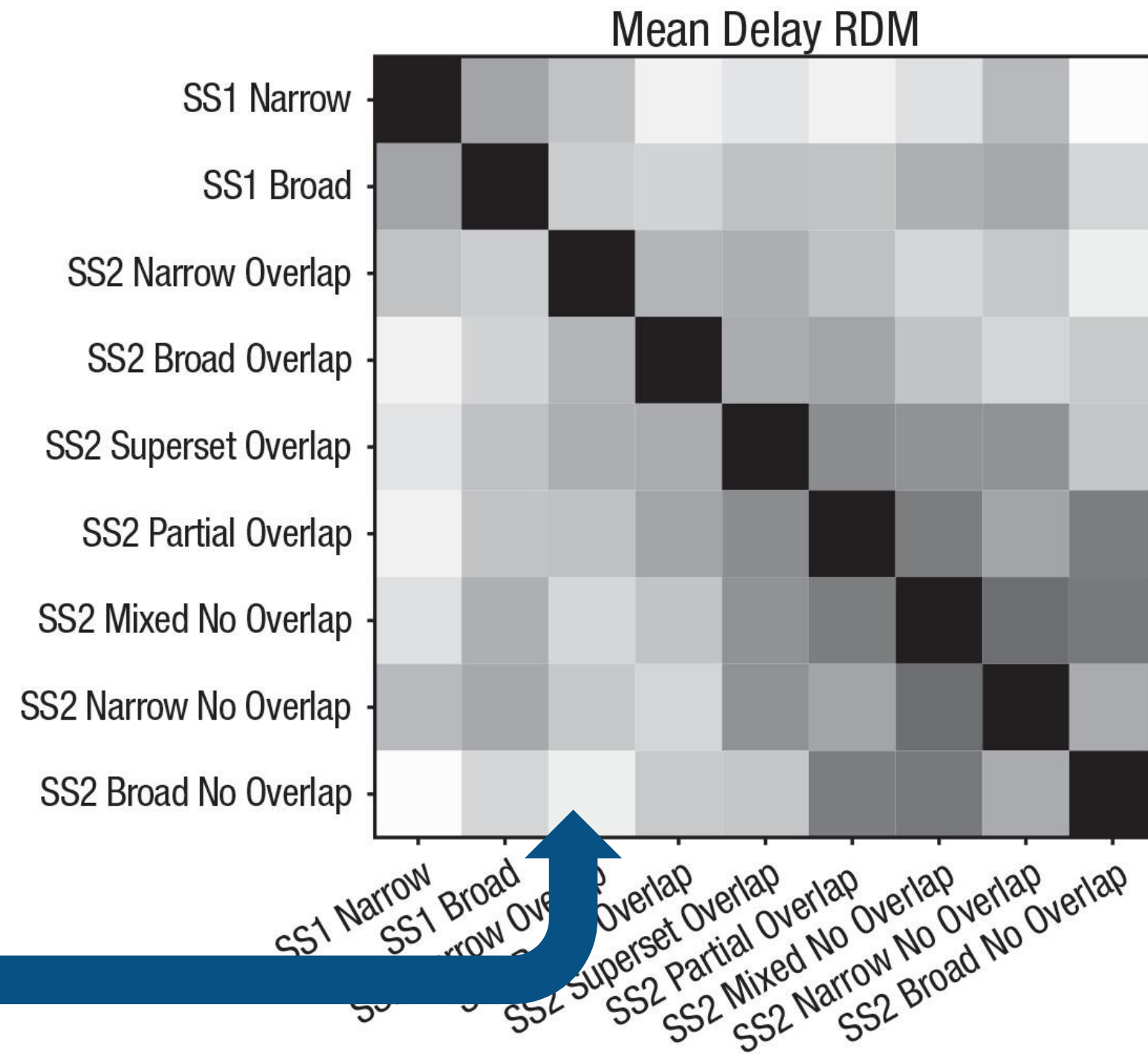


Most Dissimilar

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

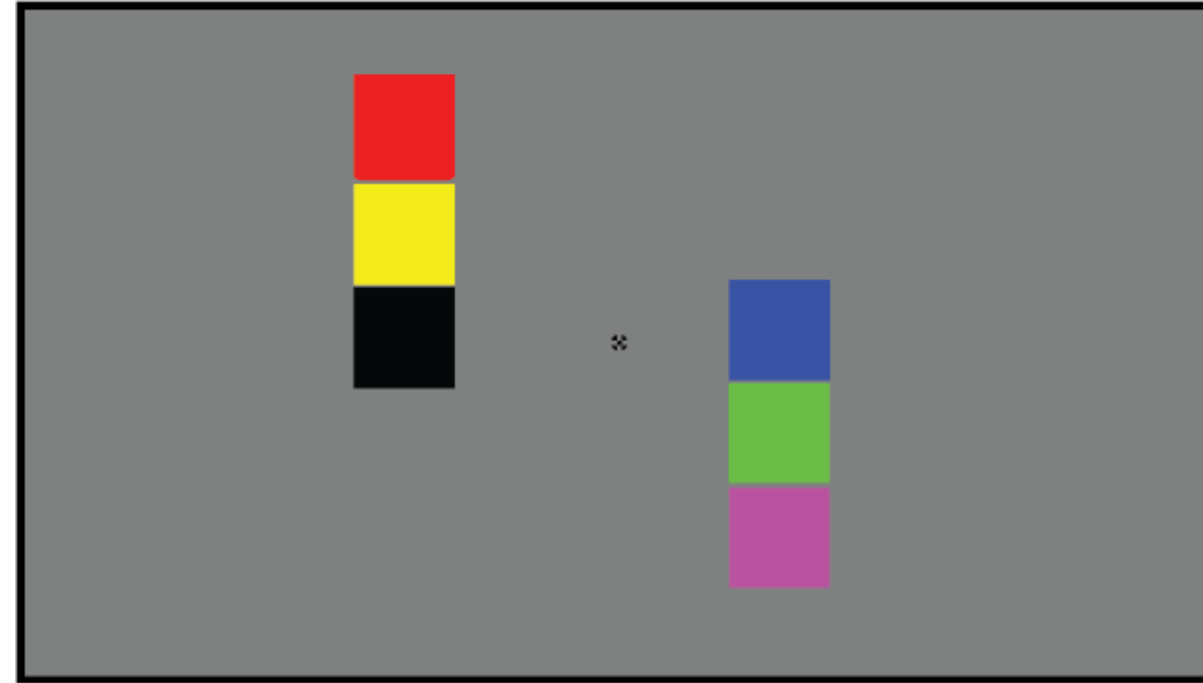
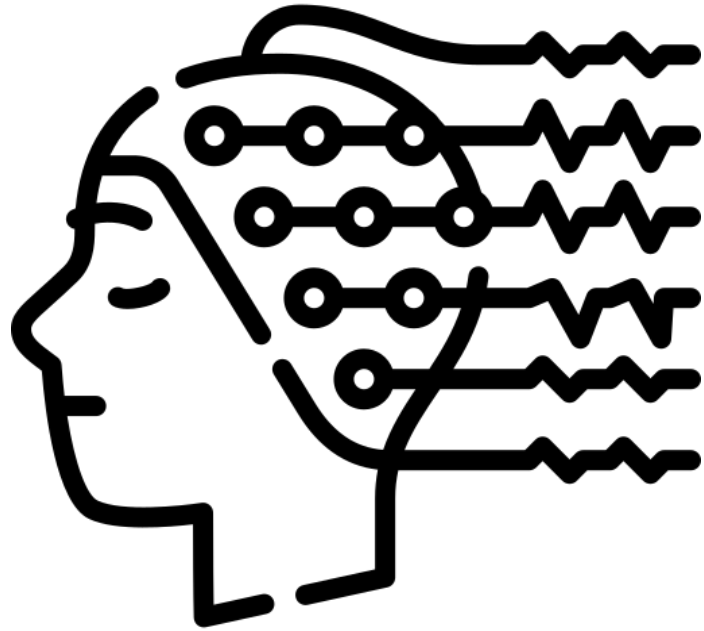
Trained subjects to learn three color triplets





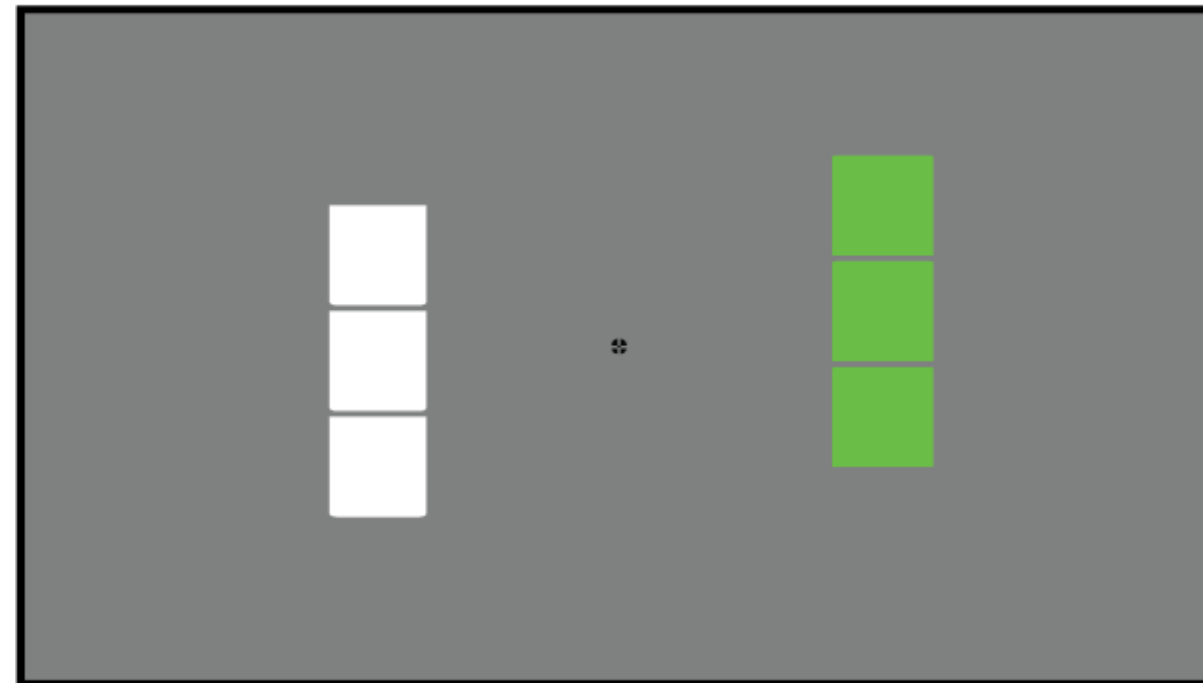


EEG



Six random
Six chunked

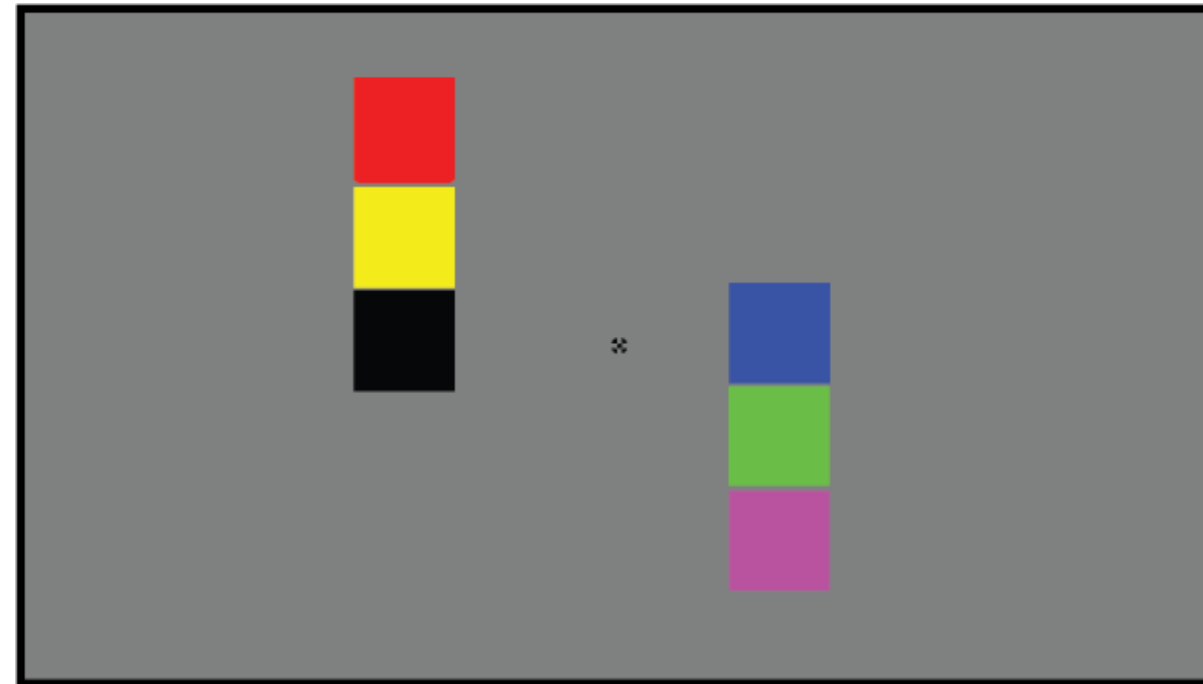
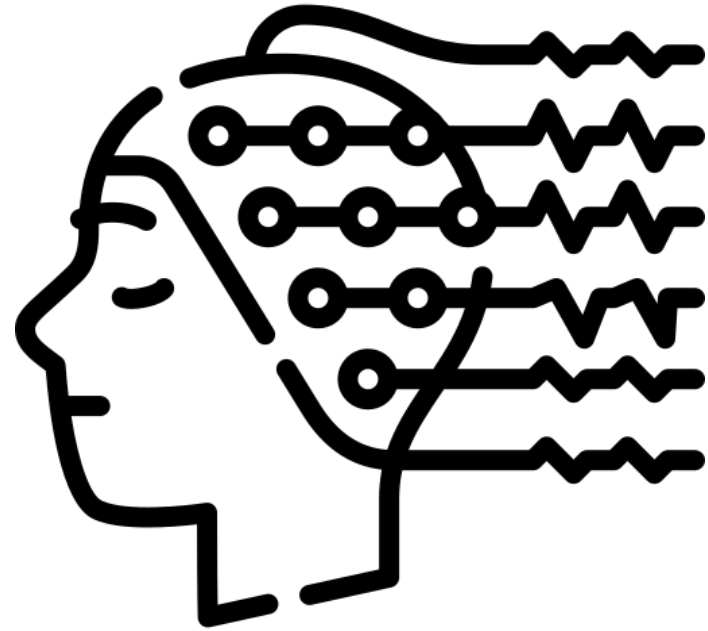
Perceptually
equivalent



Two random

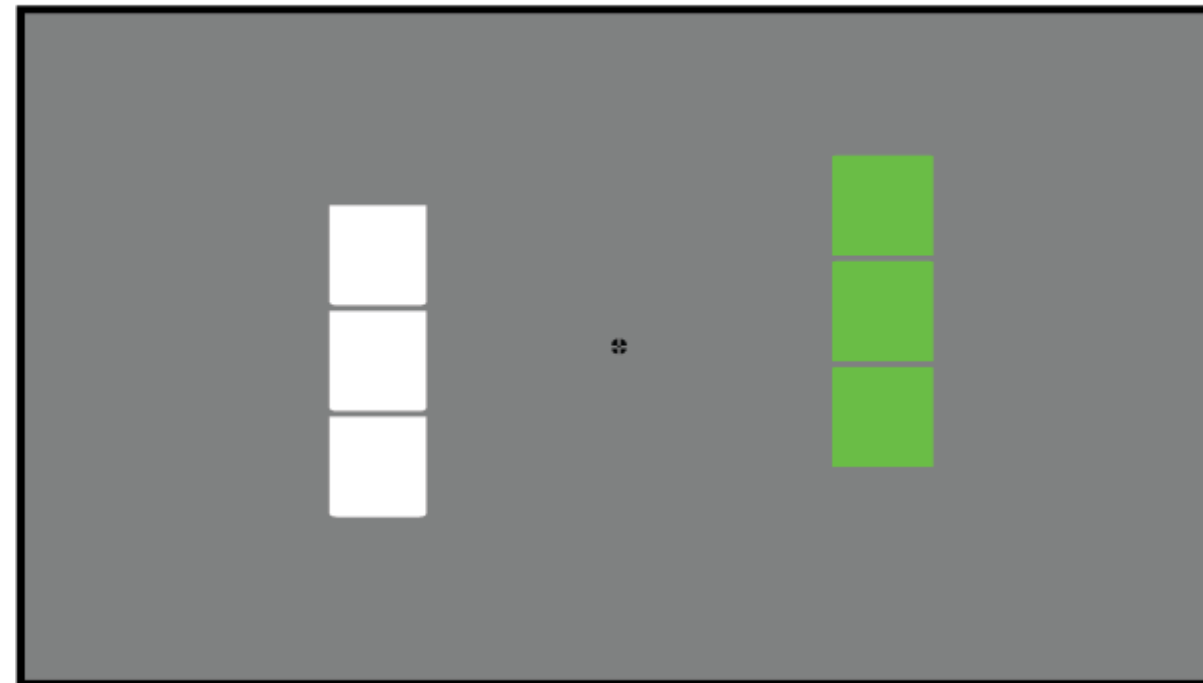


EEG



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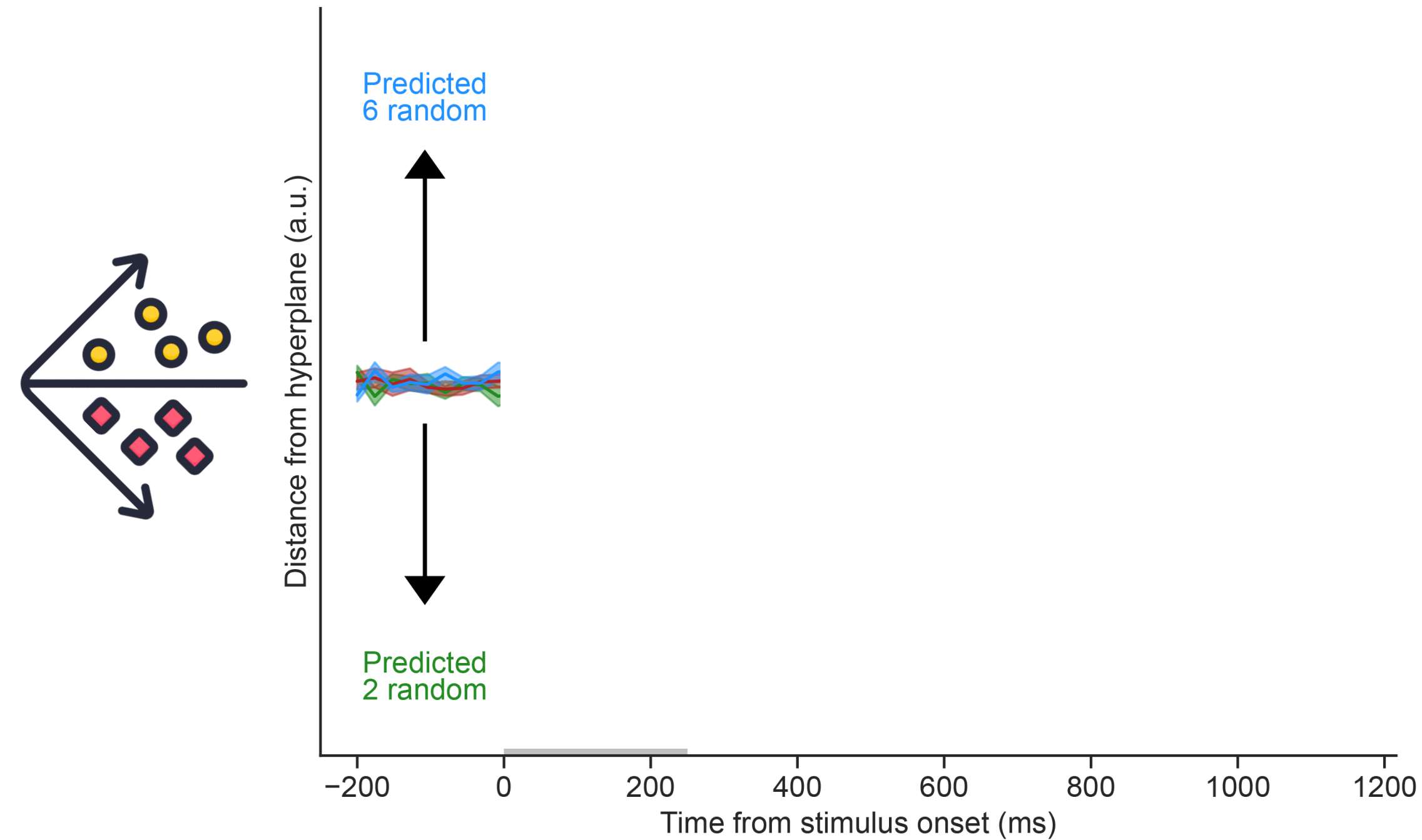


Two random

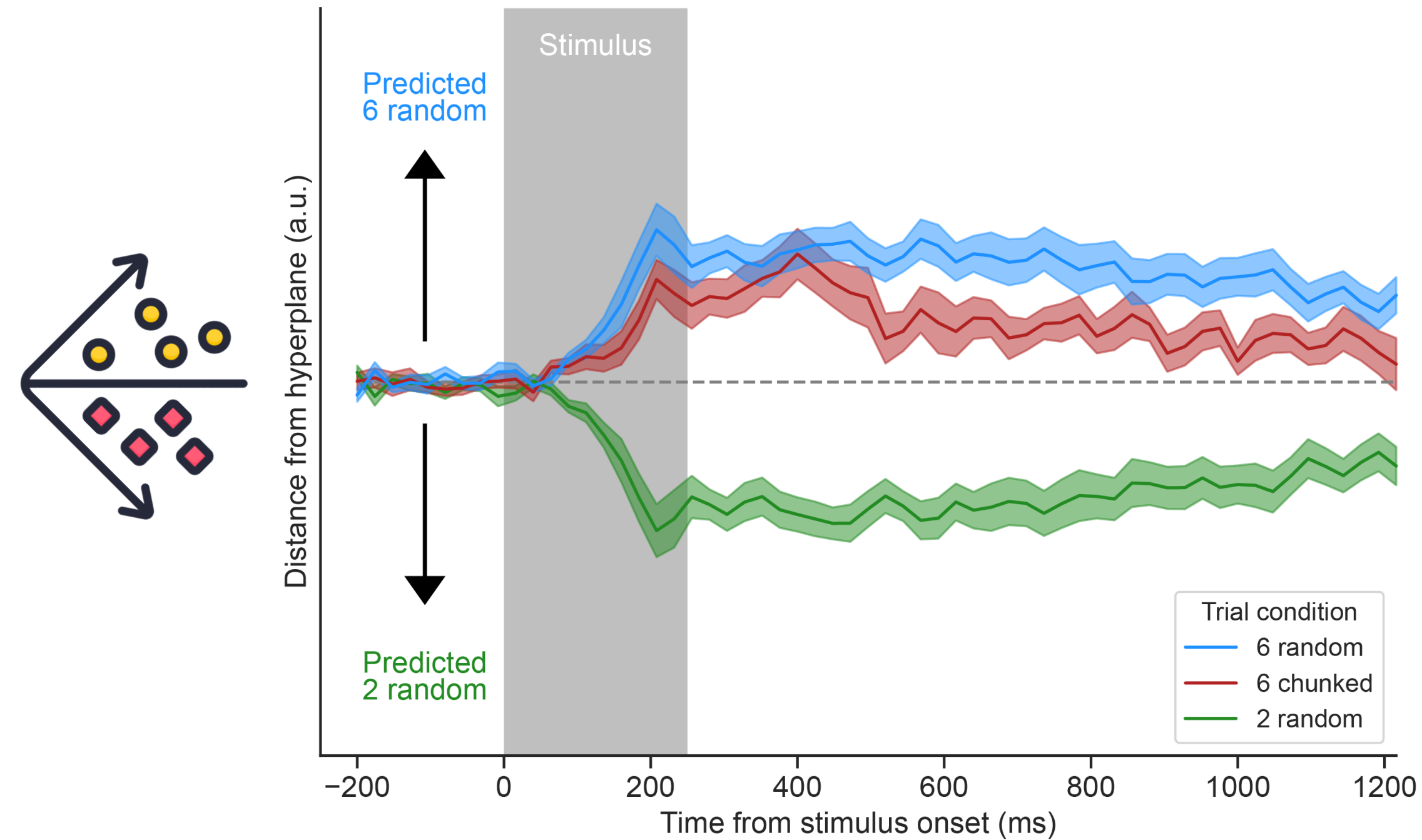


Expectation: “chunking” results in a reduction of item-based load that should be reflected in neural representations

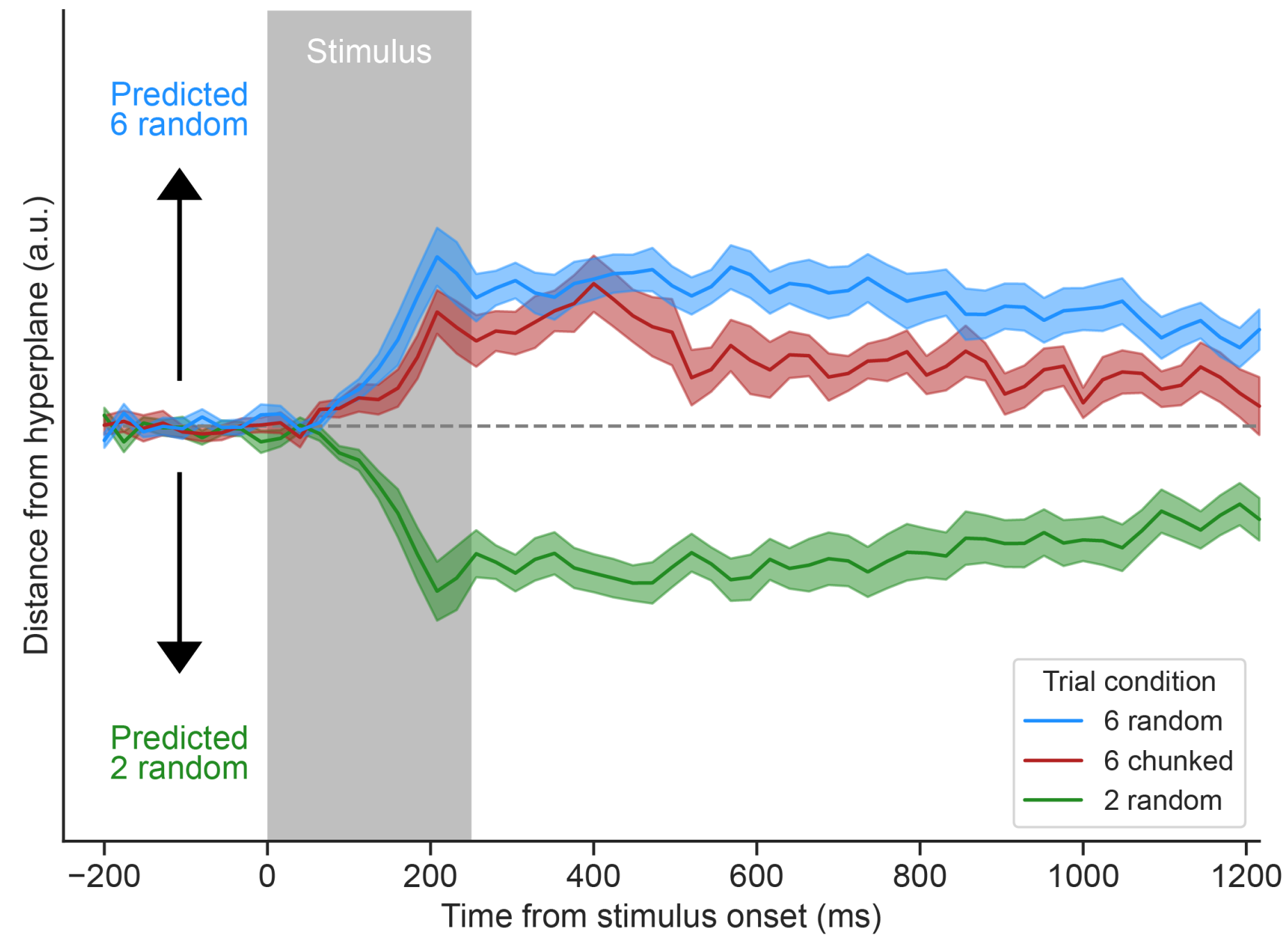
Train 6 random versus 2 random, test 6 chunked



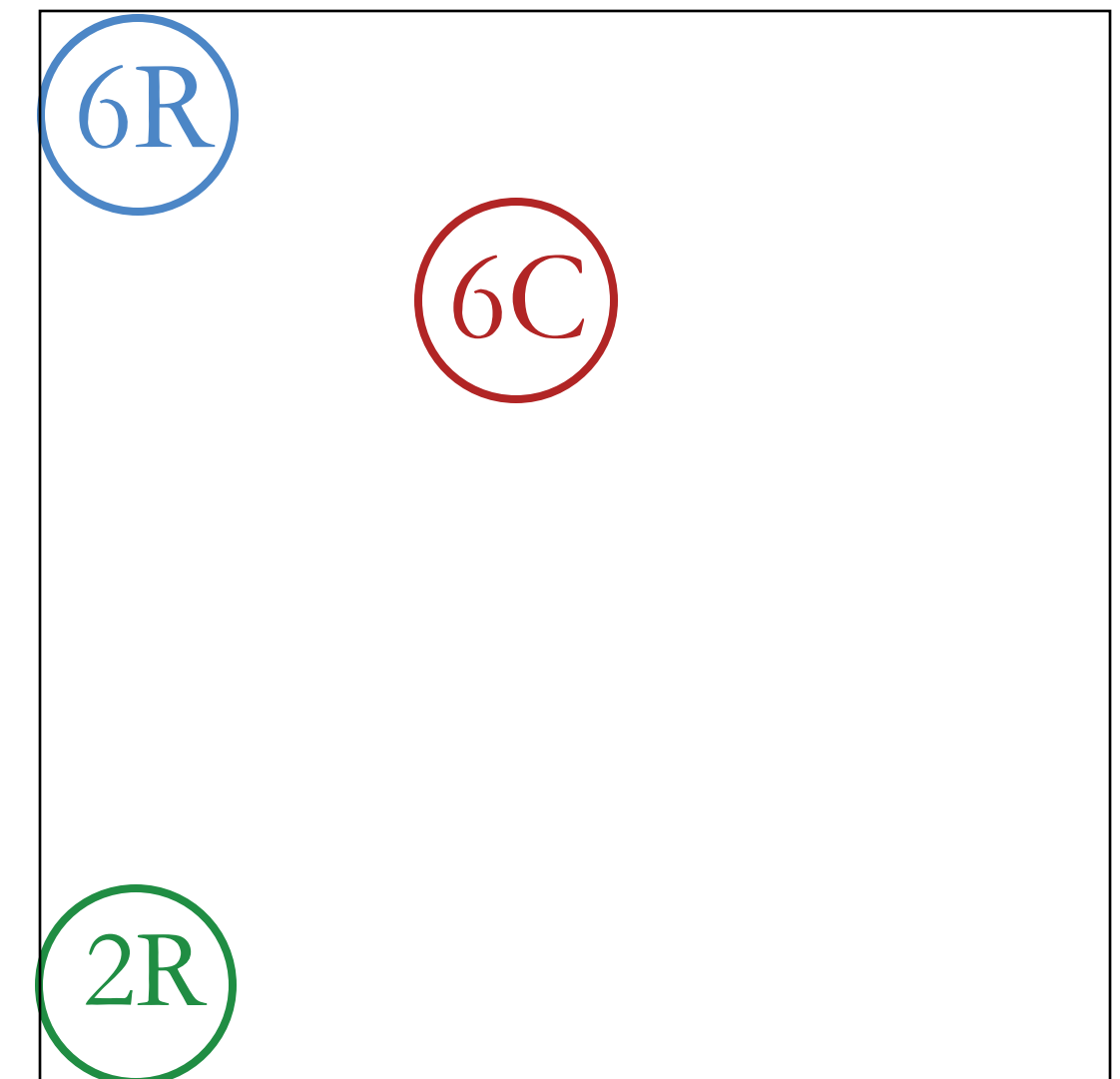
Train 6 random versus 2 random, test 6 chunked



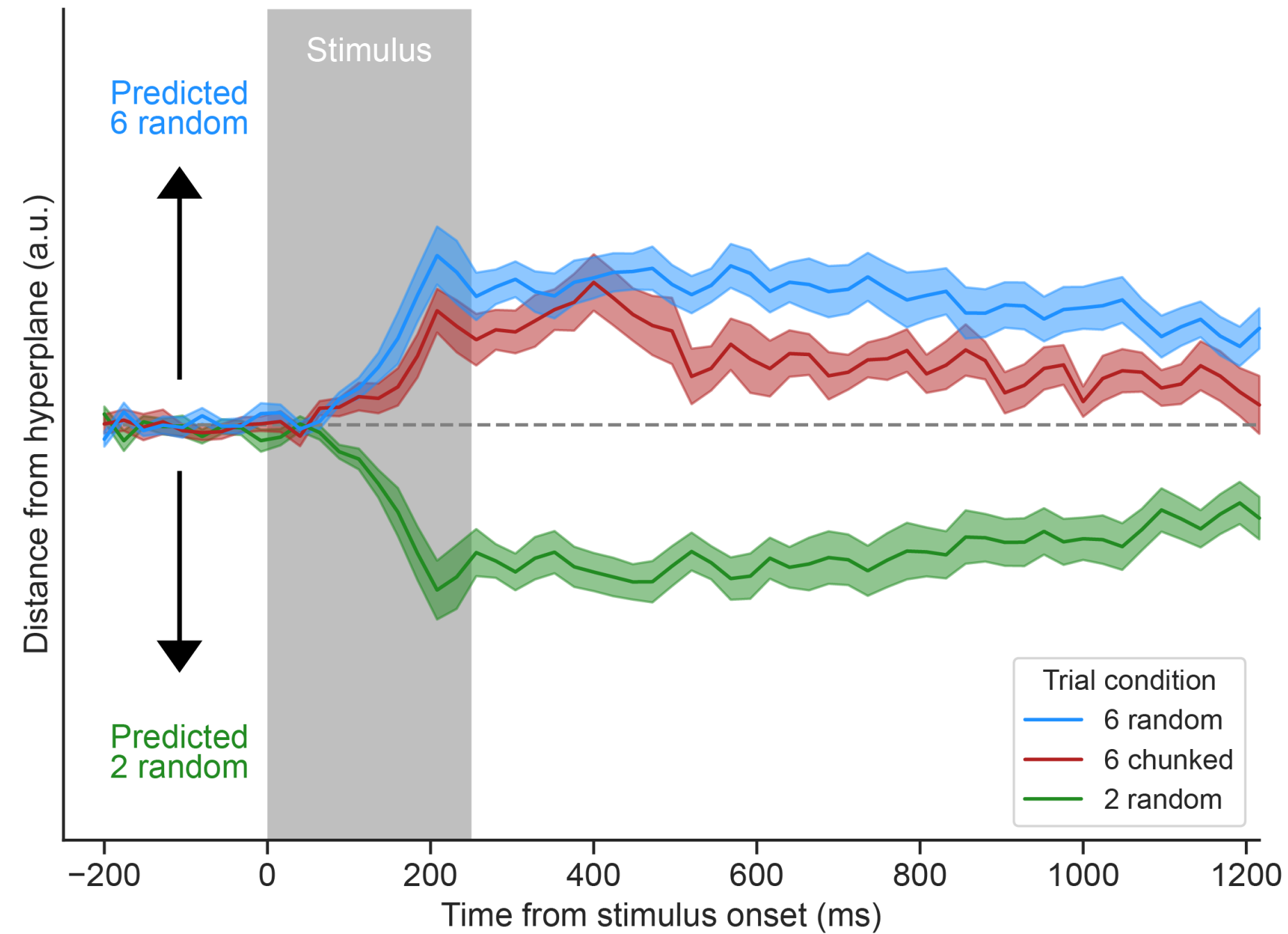
Train 6 random versus 2 random, test 6 chunked



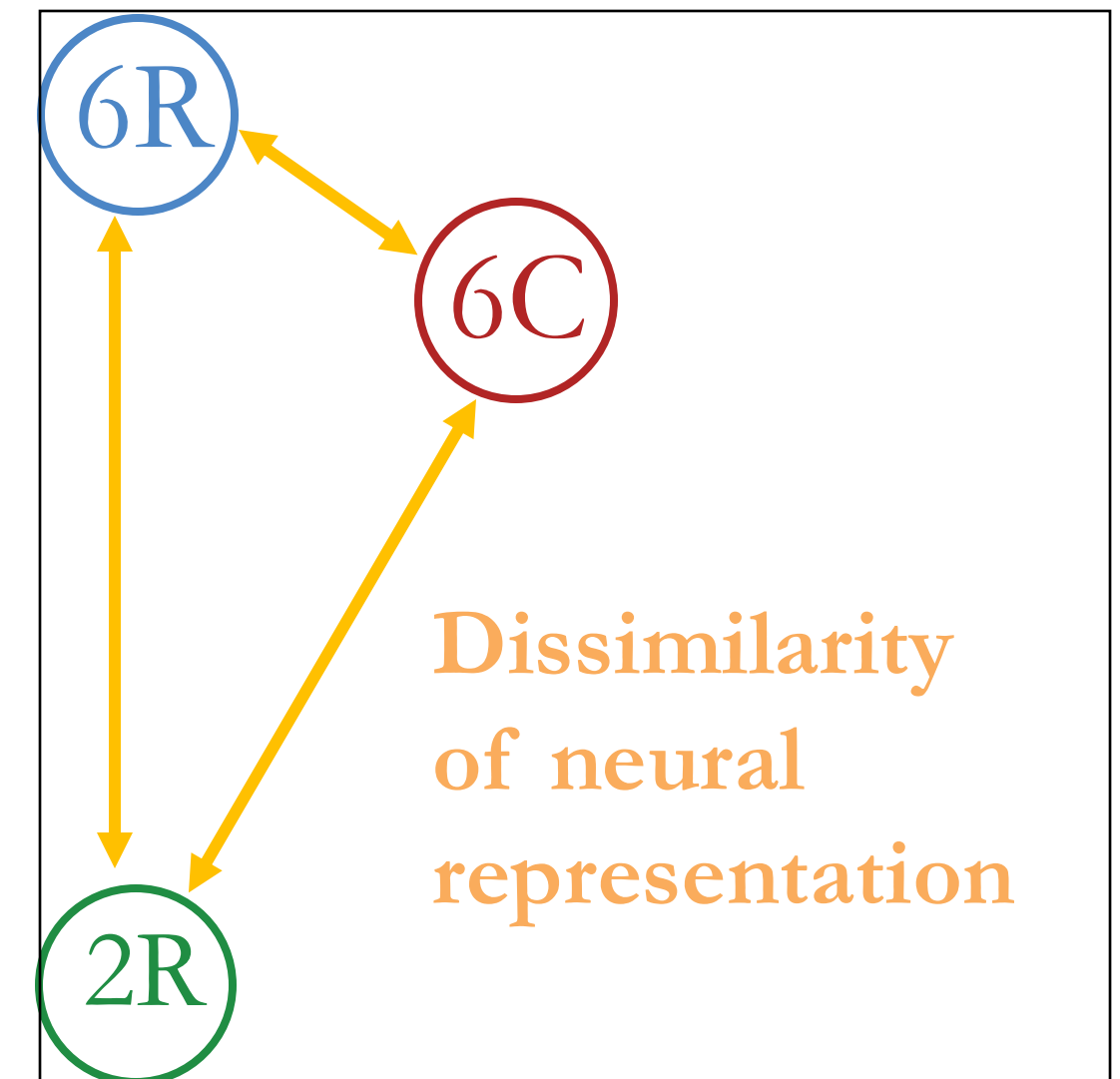
Multidimensional scaling



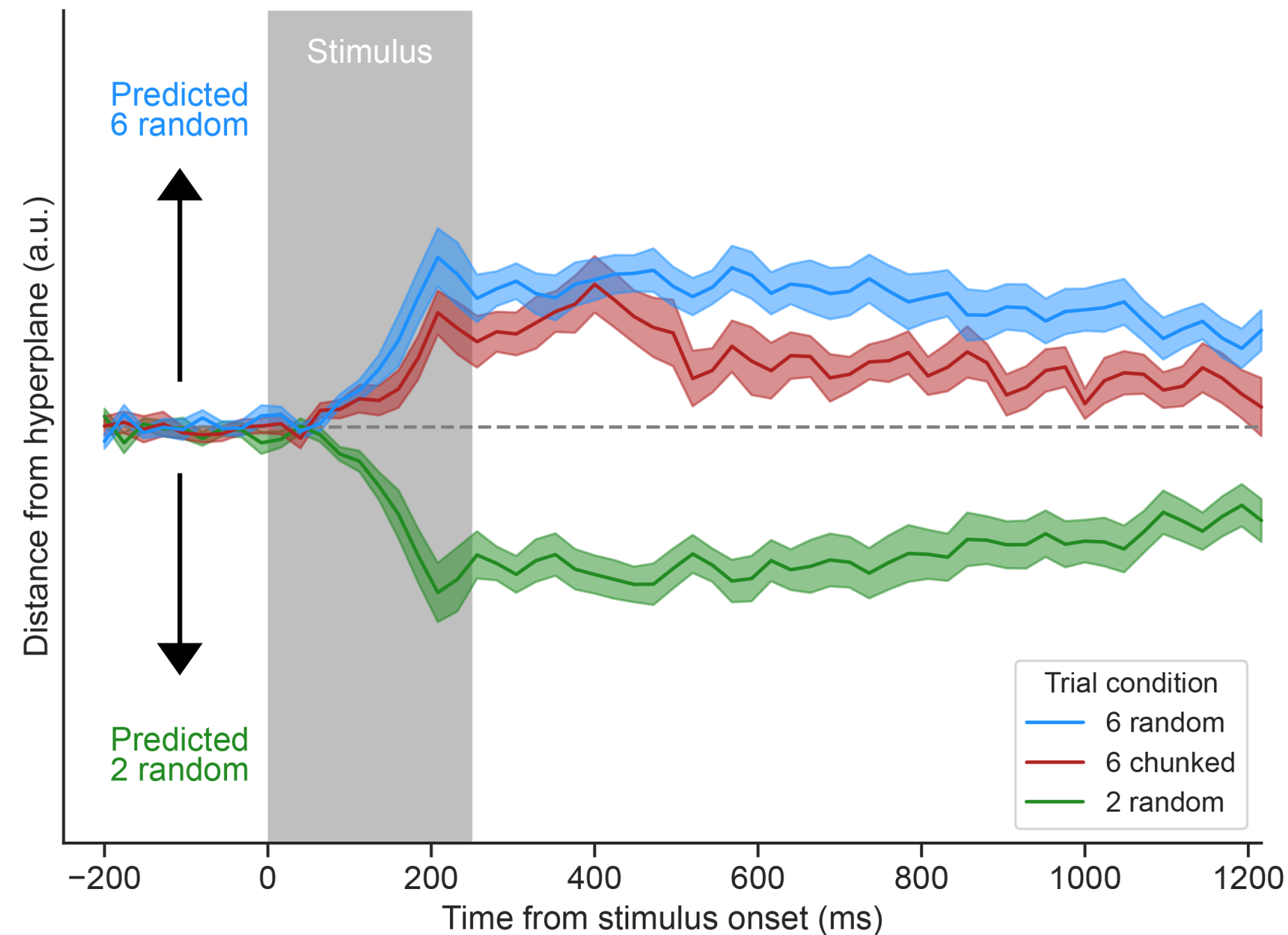
Train 6 random versus 2 random, test 6 chunked



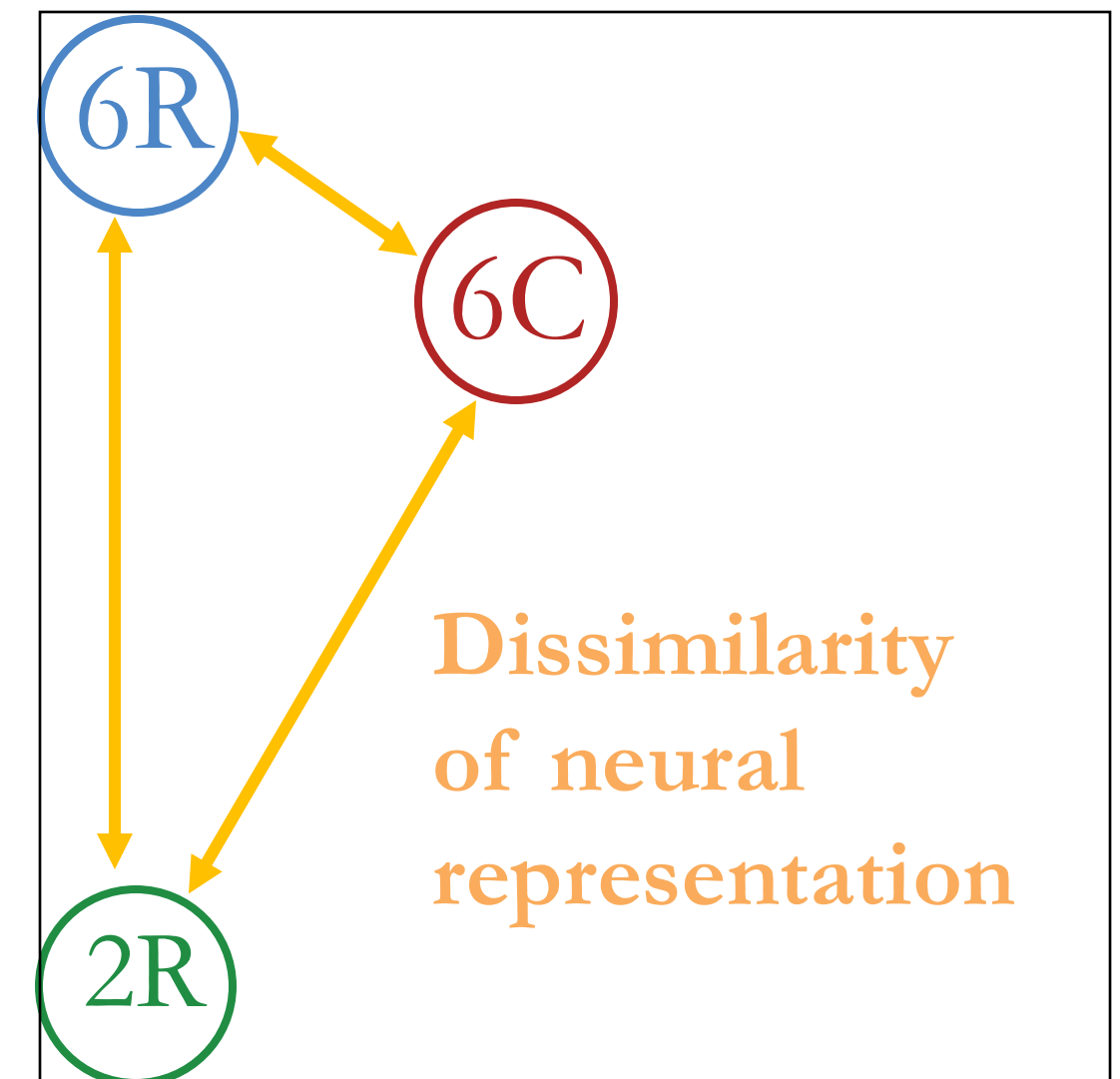
Multidimensional scaling



Train 6 random versus 2 random, test 6 chunked

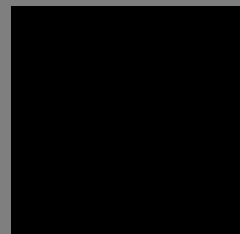
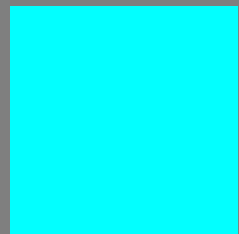
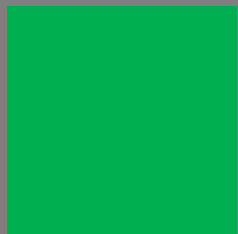
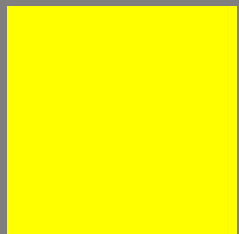
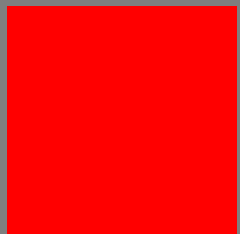
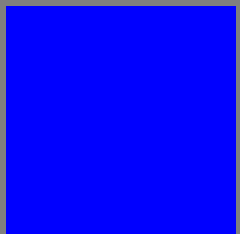


Multidimensional scaling



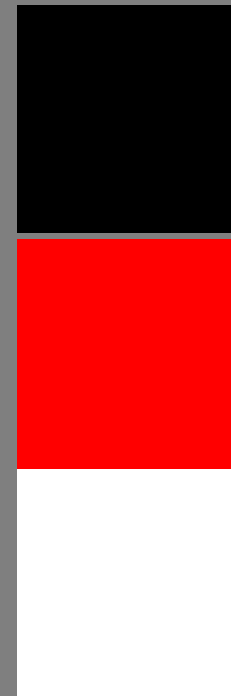
What can we infer from these results?
What cognitive model would predict or could explain this
pattern of representational similarity?

Awareness Test

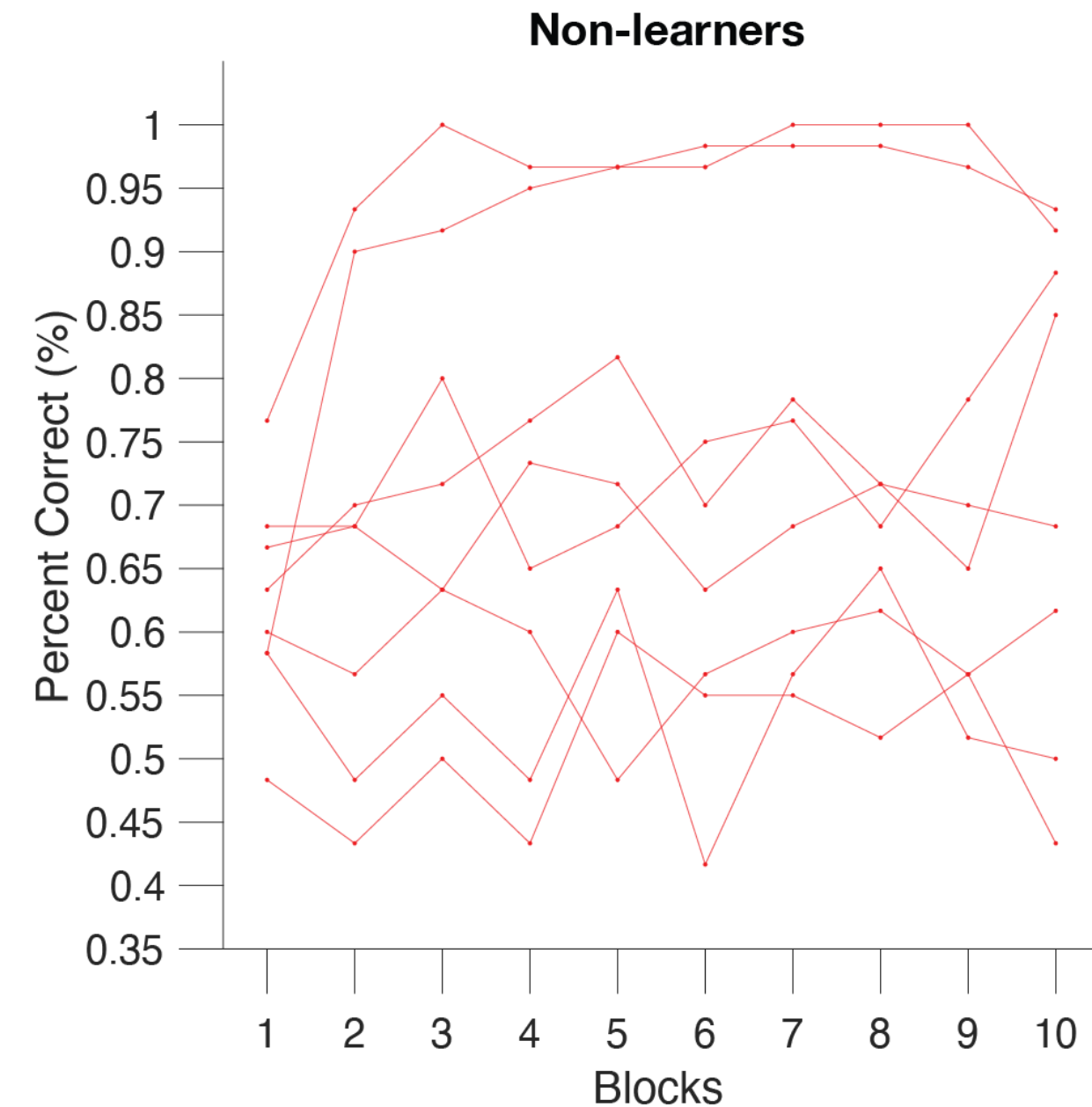
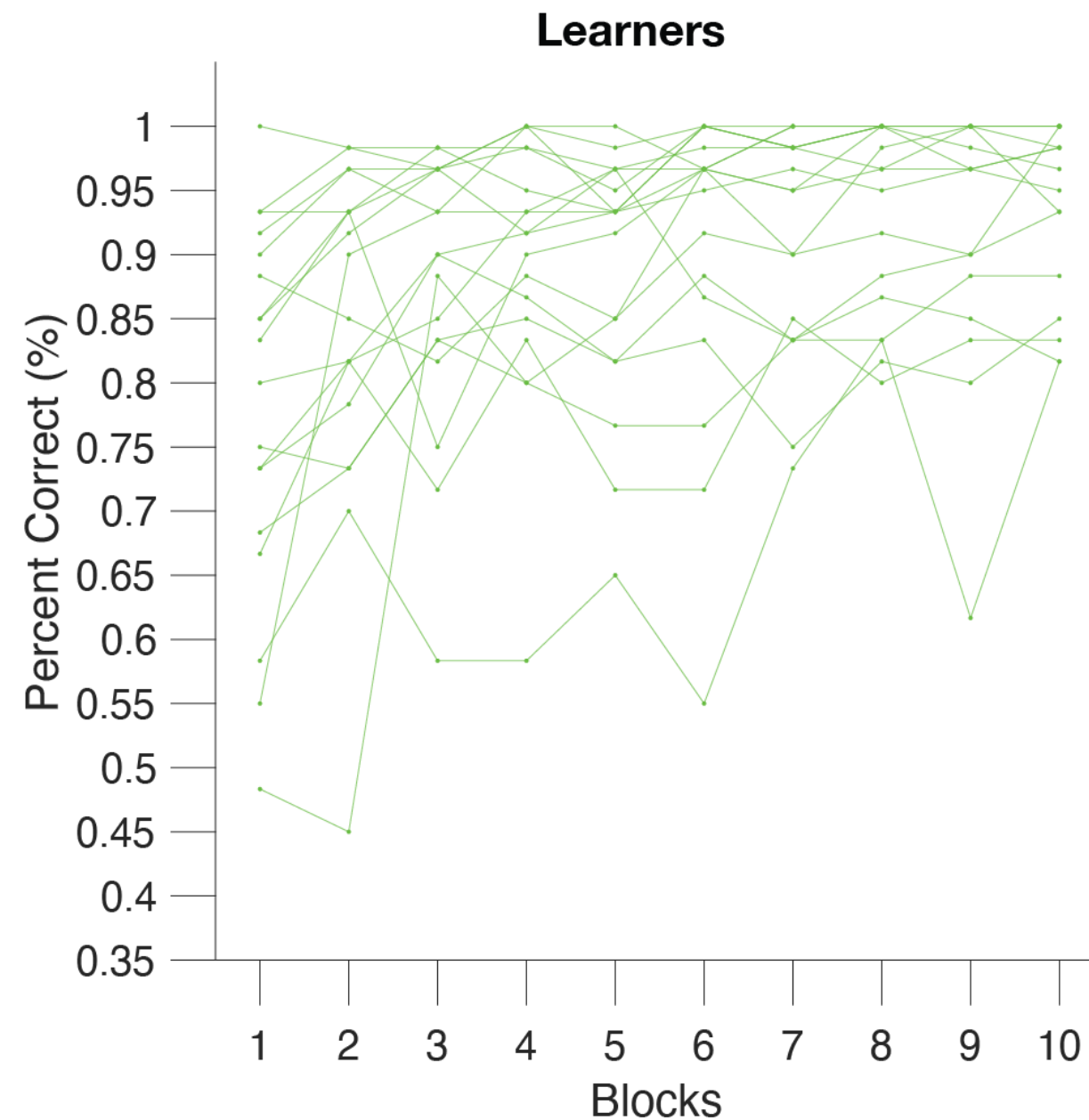
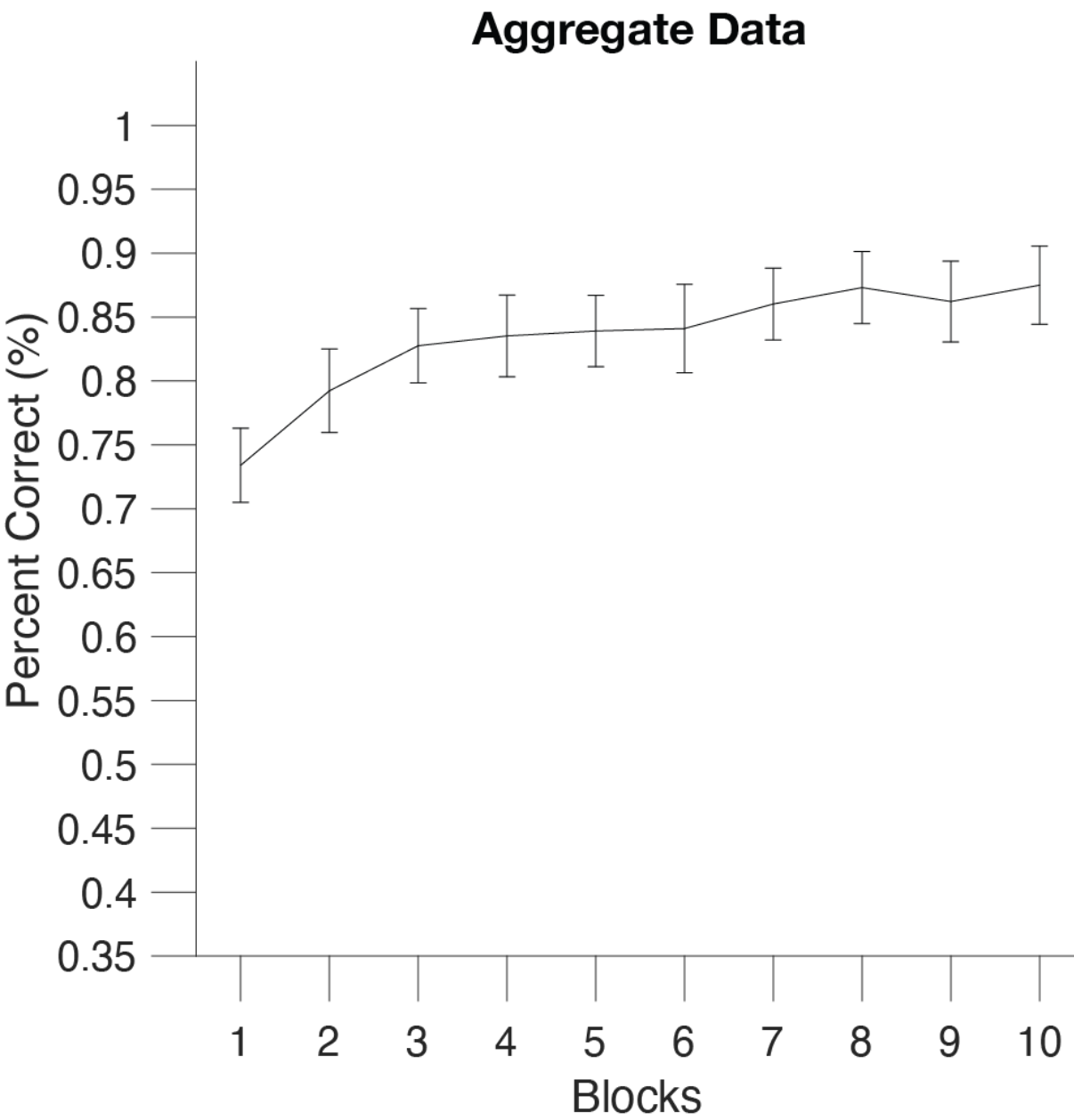


Awareness Test

Only subjects that recreated **all** triplets were considered “aware”

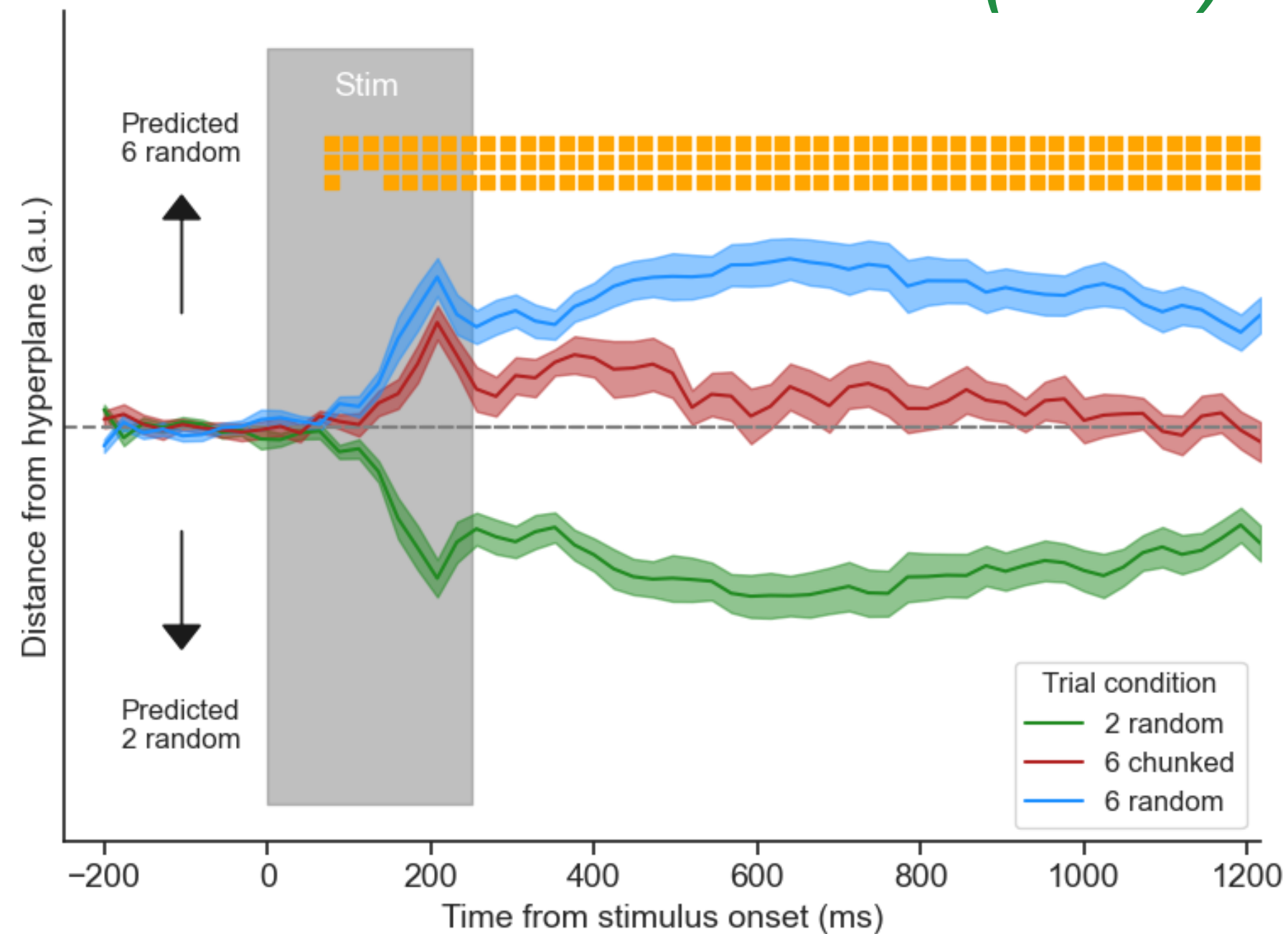


Training Results

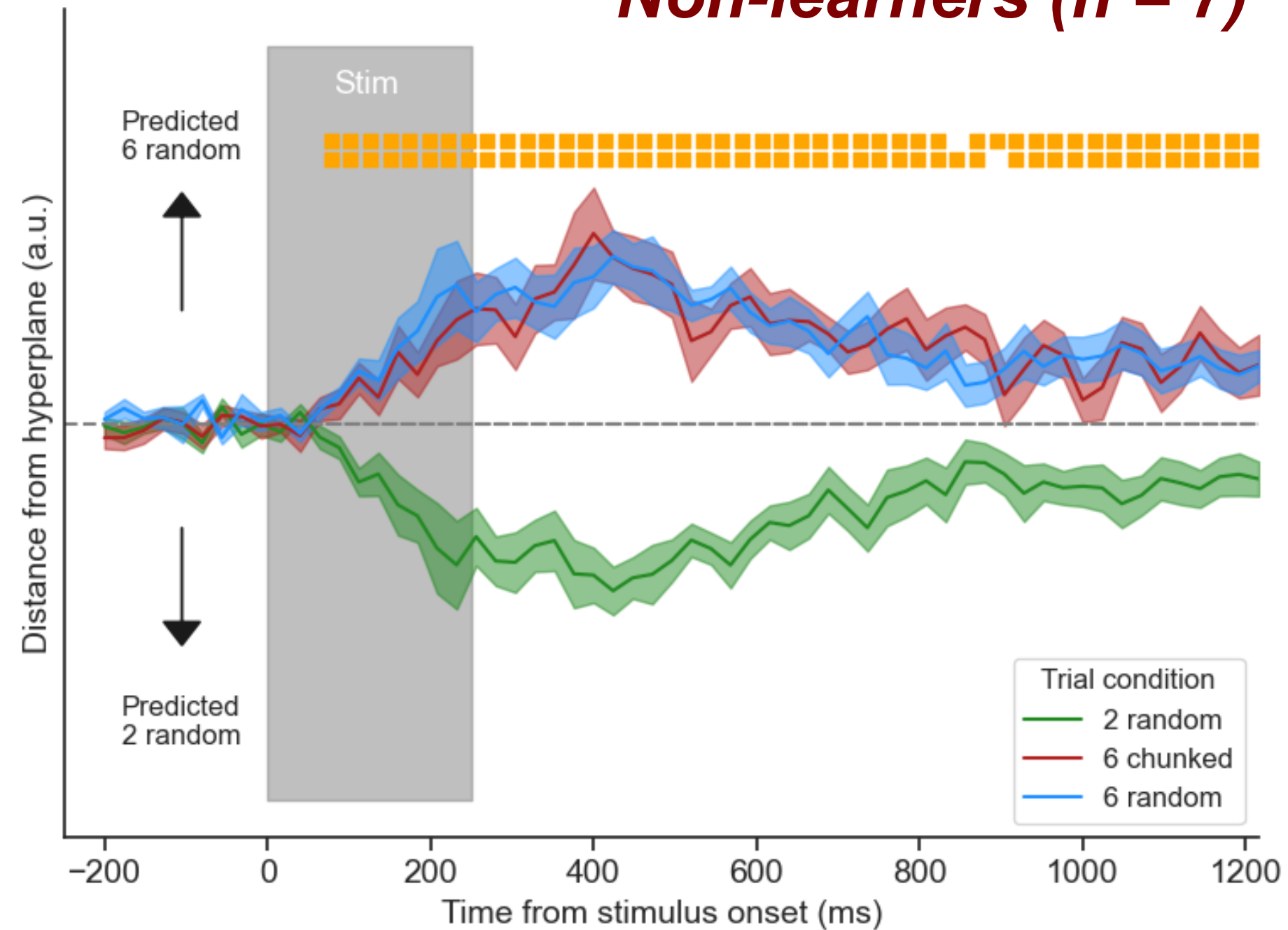


Learners vs non-learners

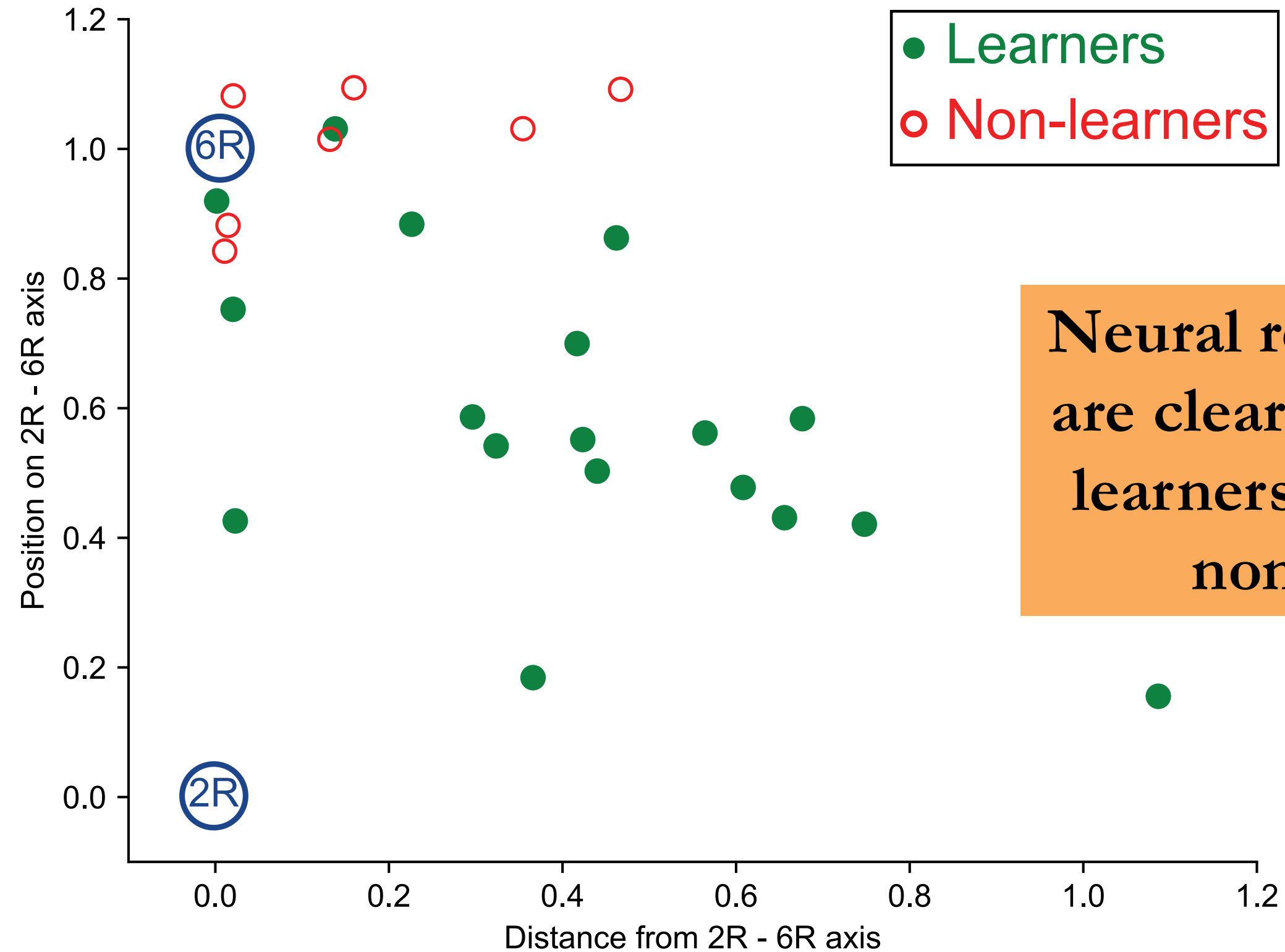
Learners (n = 18)



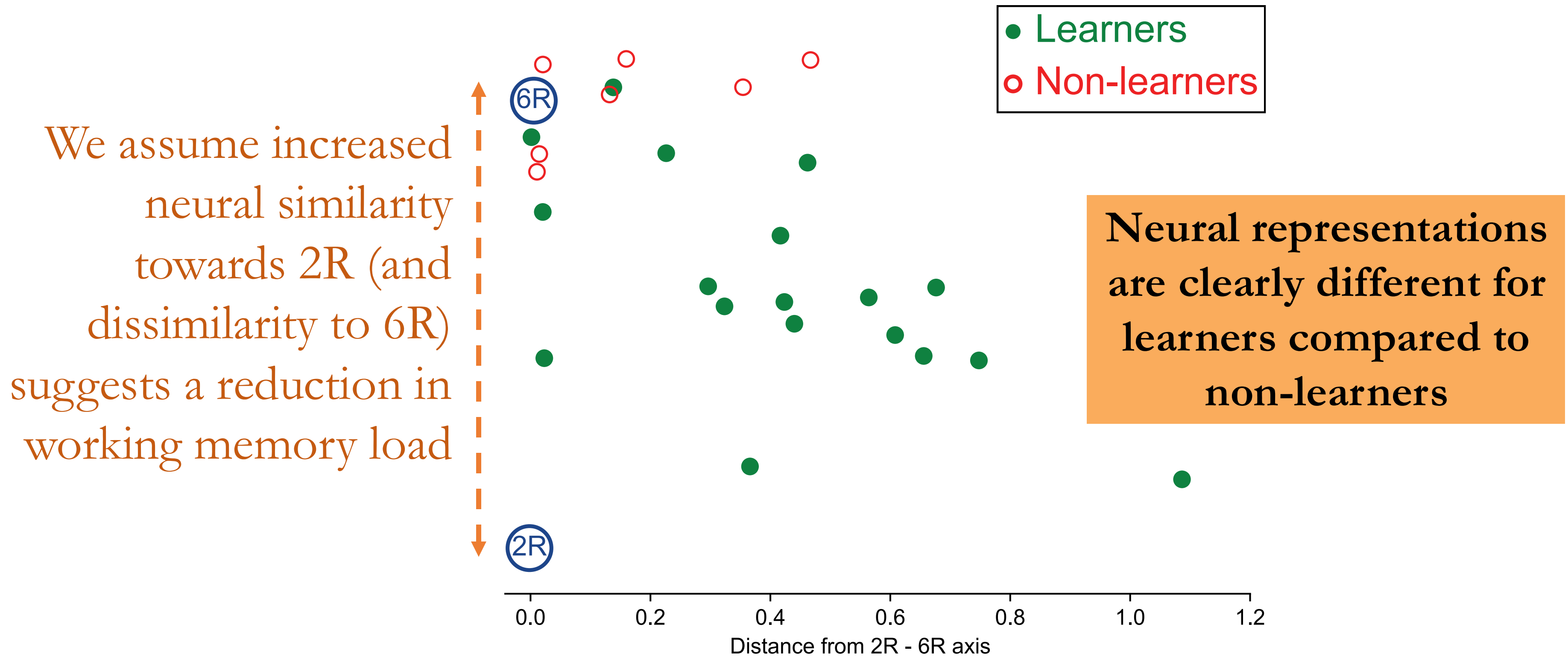
Non-learners (n = 7)



Multidimensional scaling on each subject

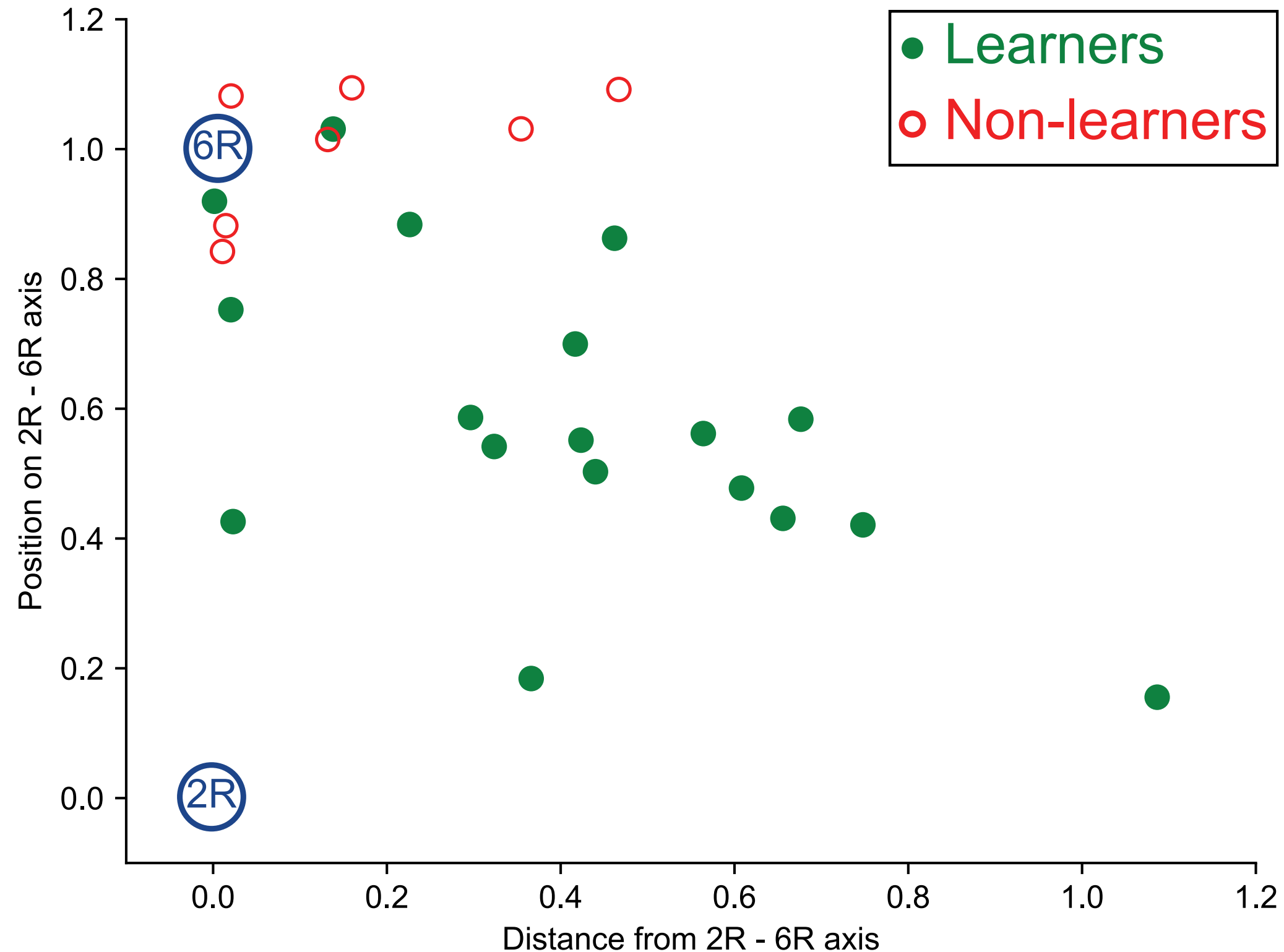


Multidimensional scaling on each subject



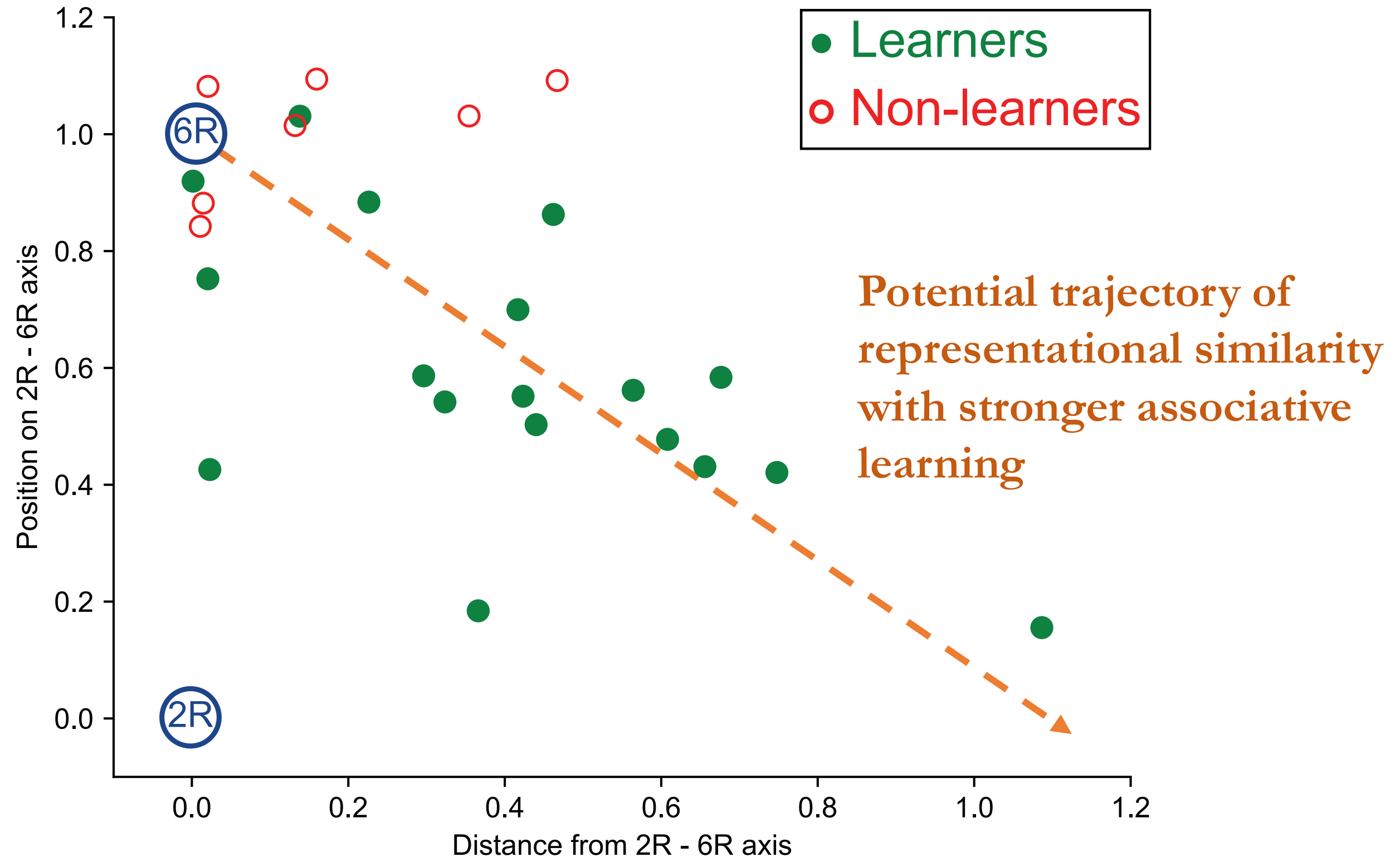
Multidimensional scaling on each subject

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**.



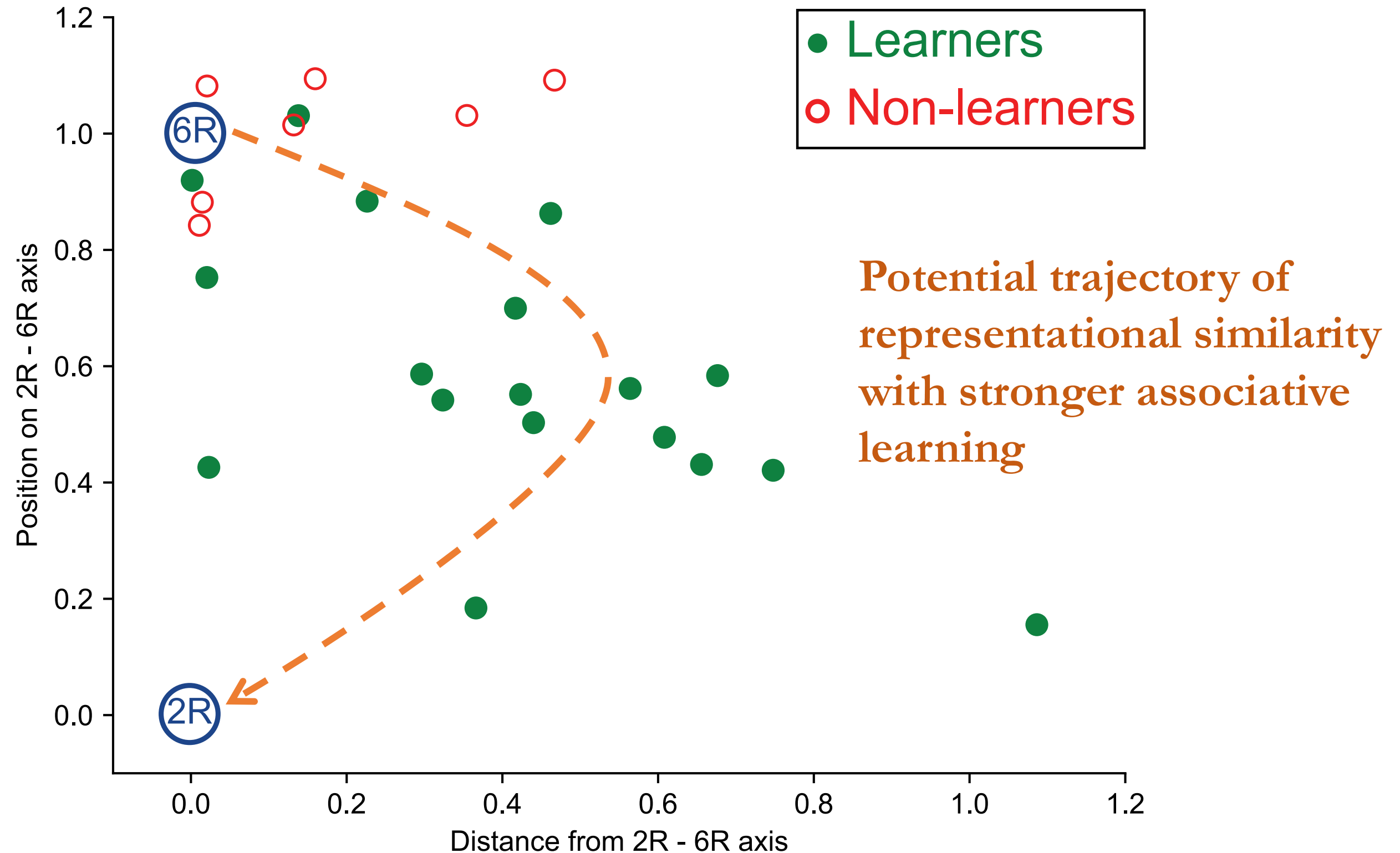
Multidimensional scaling on each subject

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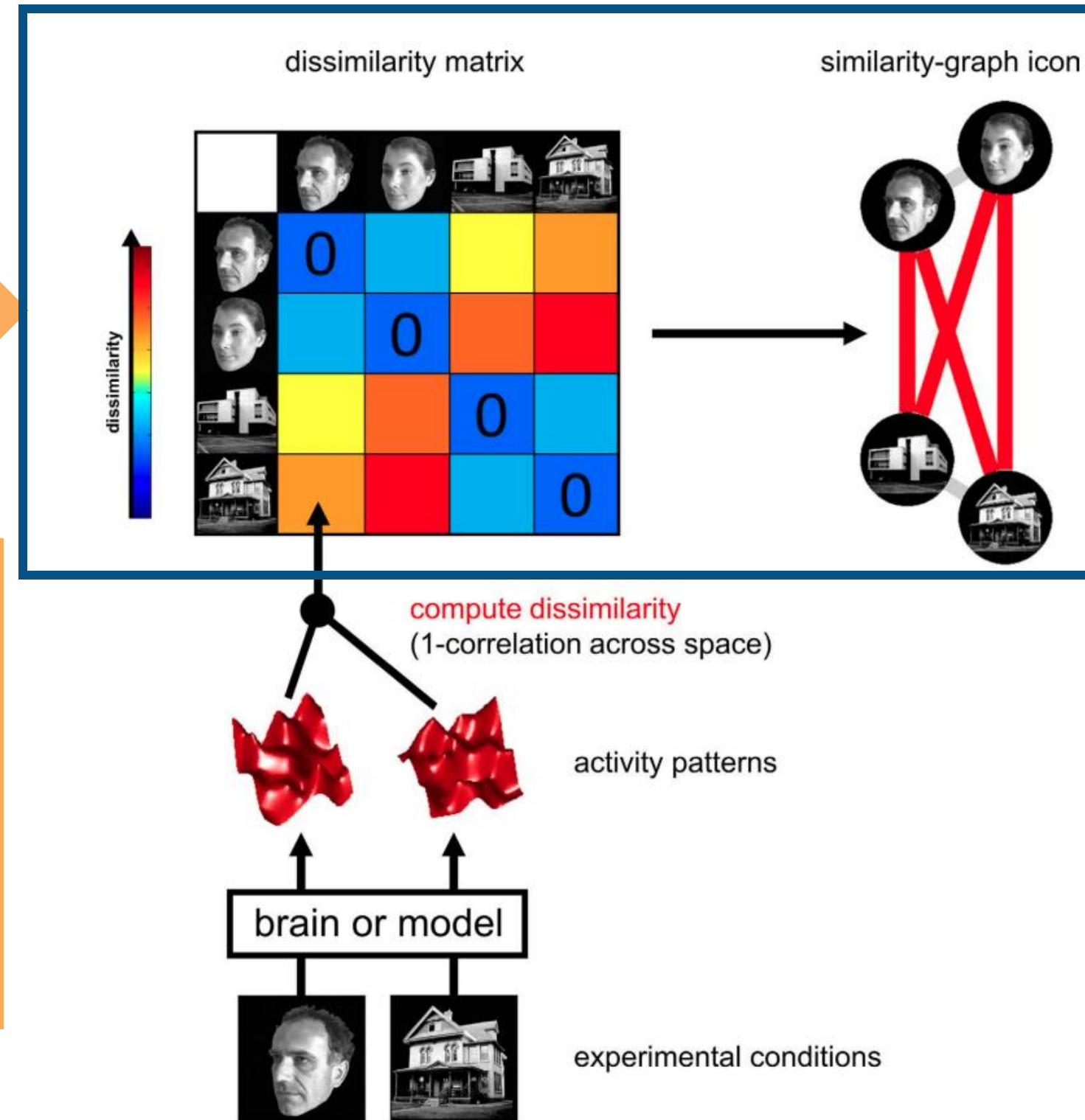


Multidimensional scaling on each subject

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.

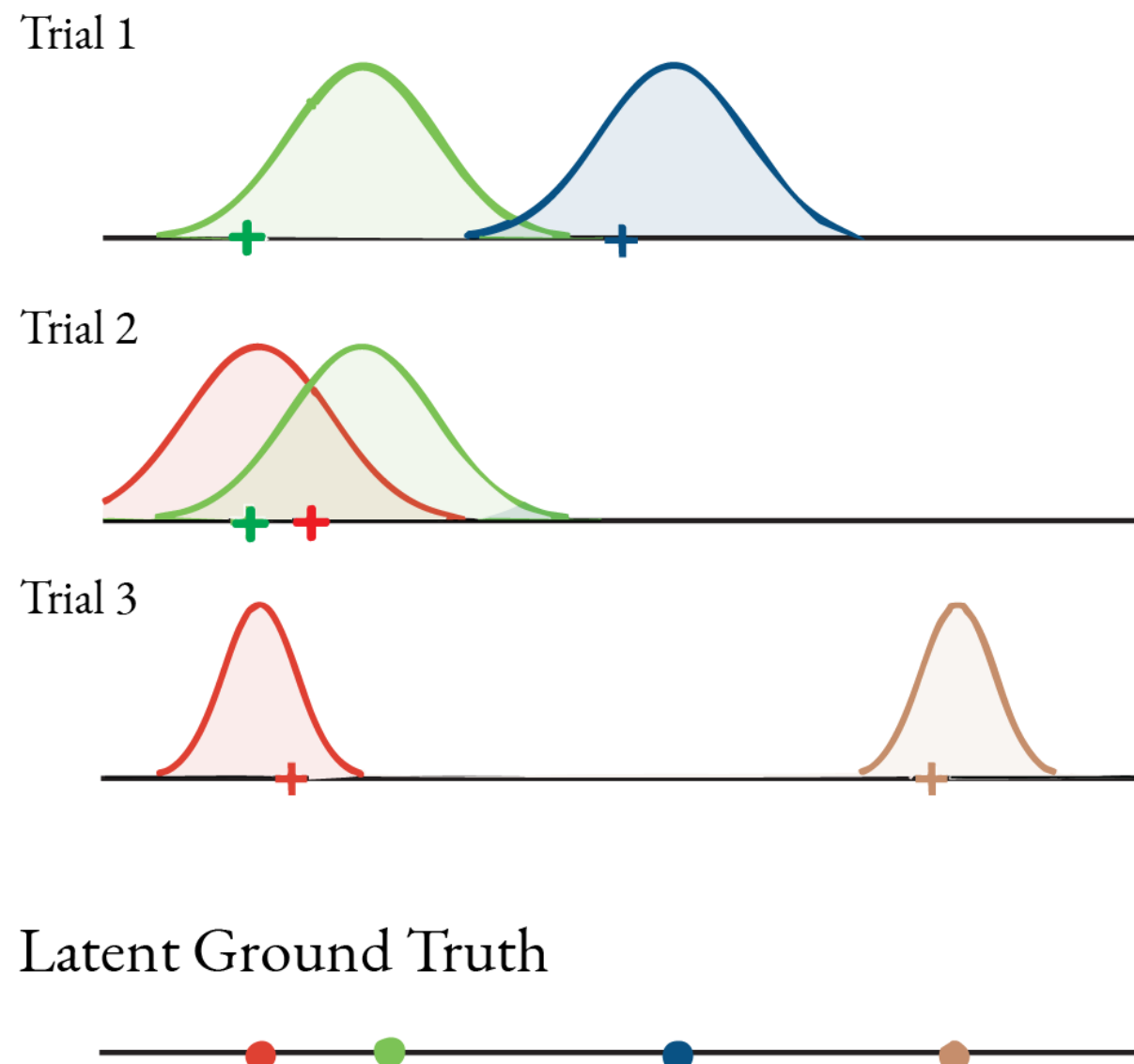
Representations in formal cognitive models

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

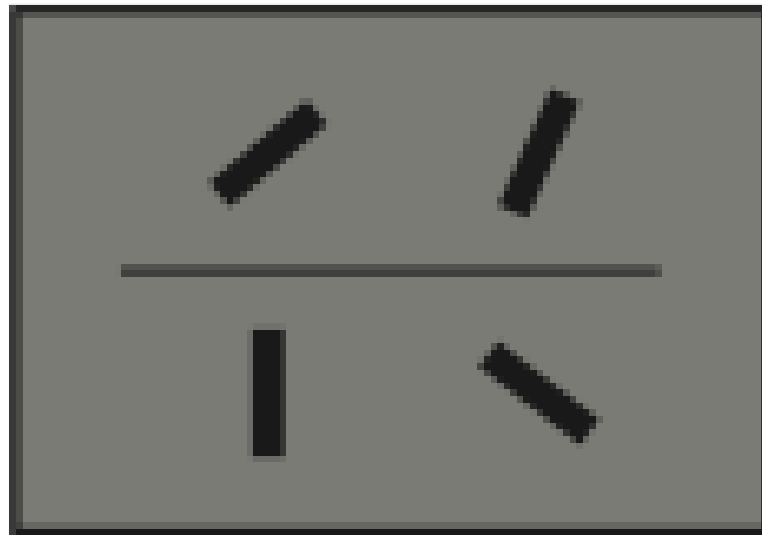
In brief, we used Bayesian MCMC methods to recover the **latent representation of oriented lines** used in three cognitive tasks.



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

The cognitive tasks

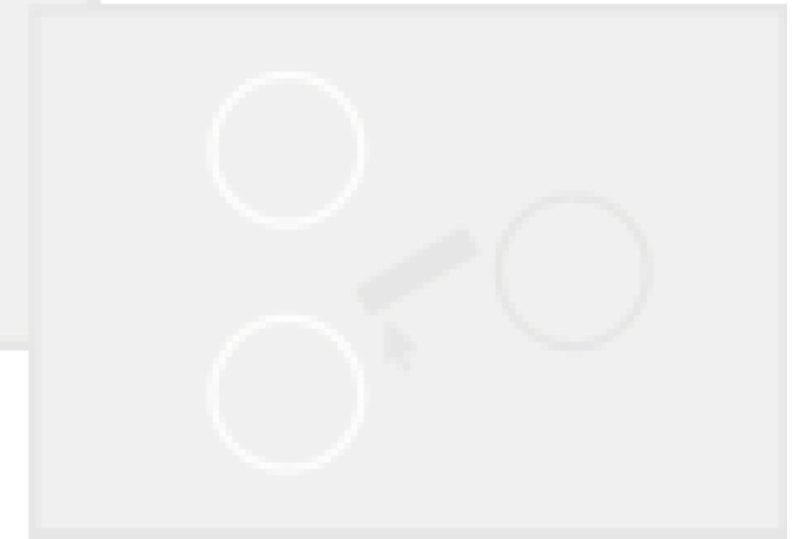
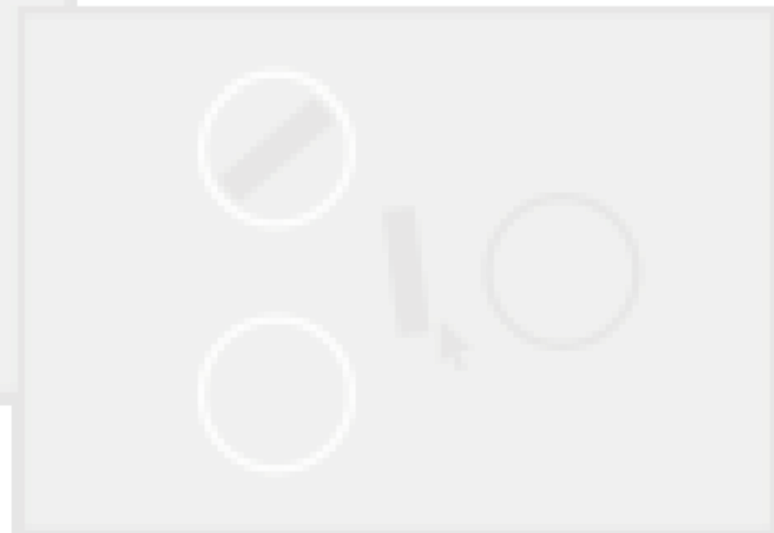
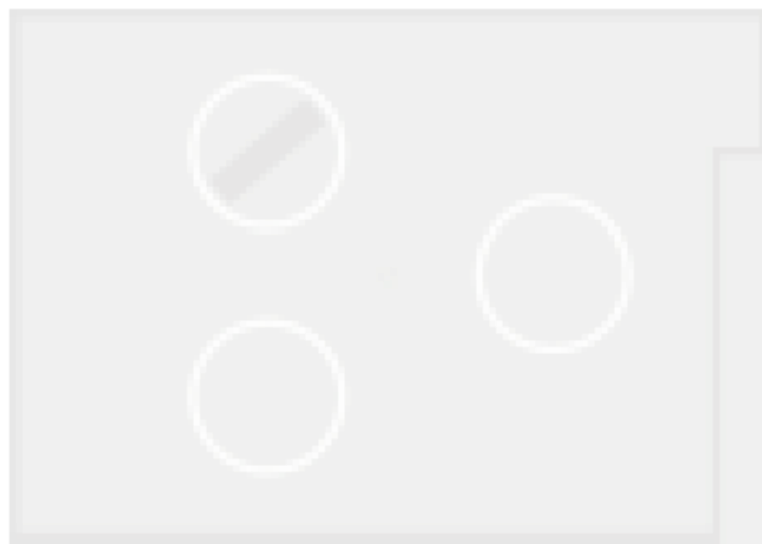
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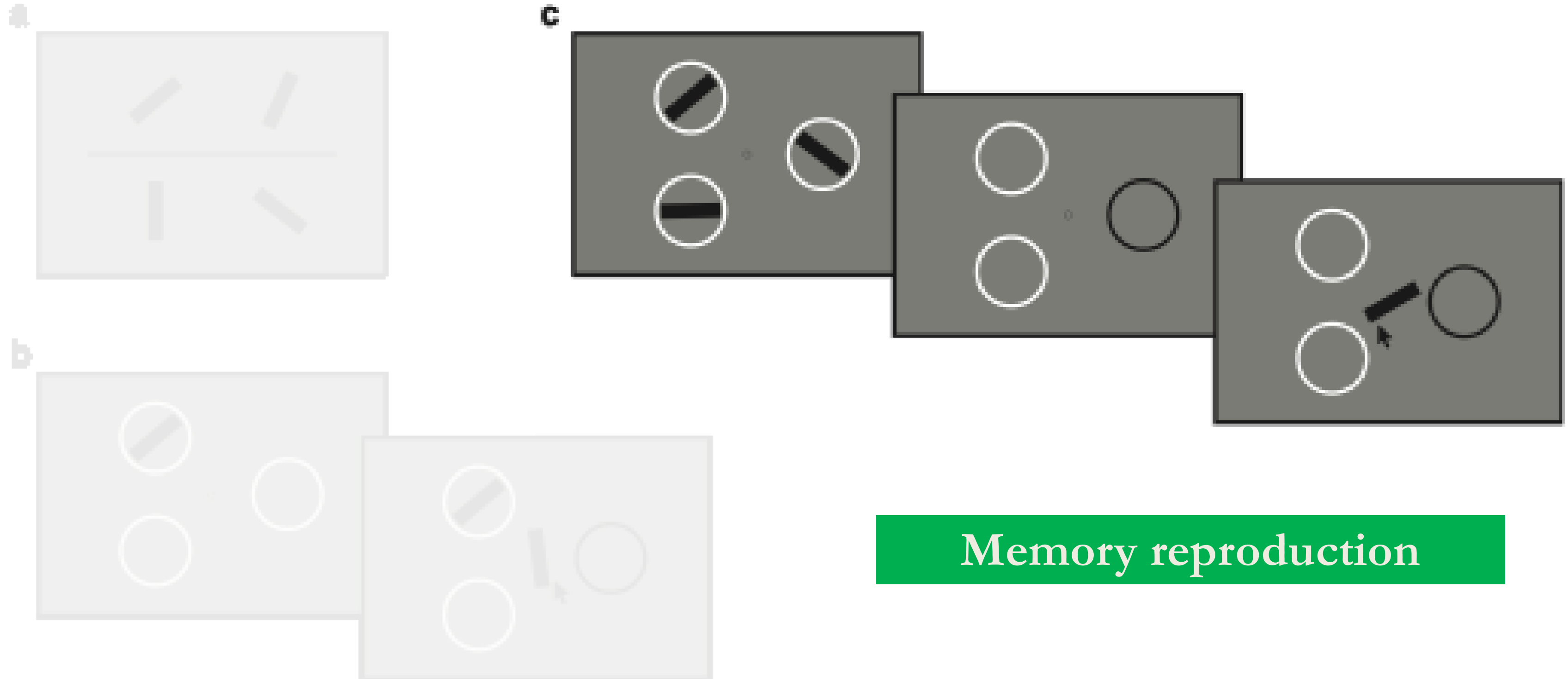
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Similarity comparison

b



The cognitive tasks

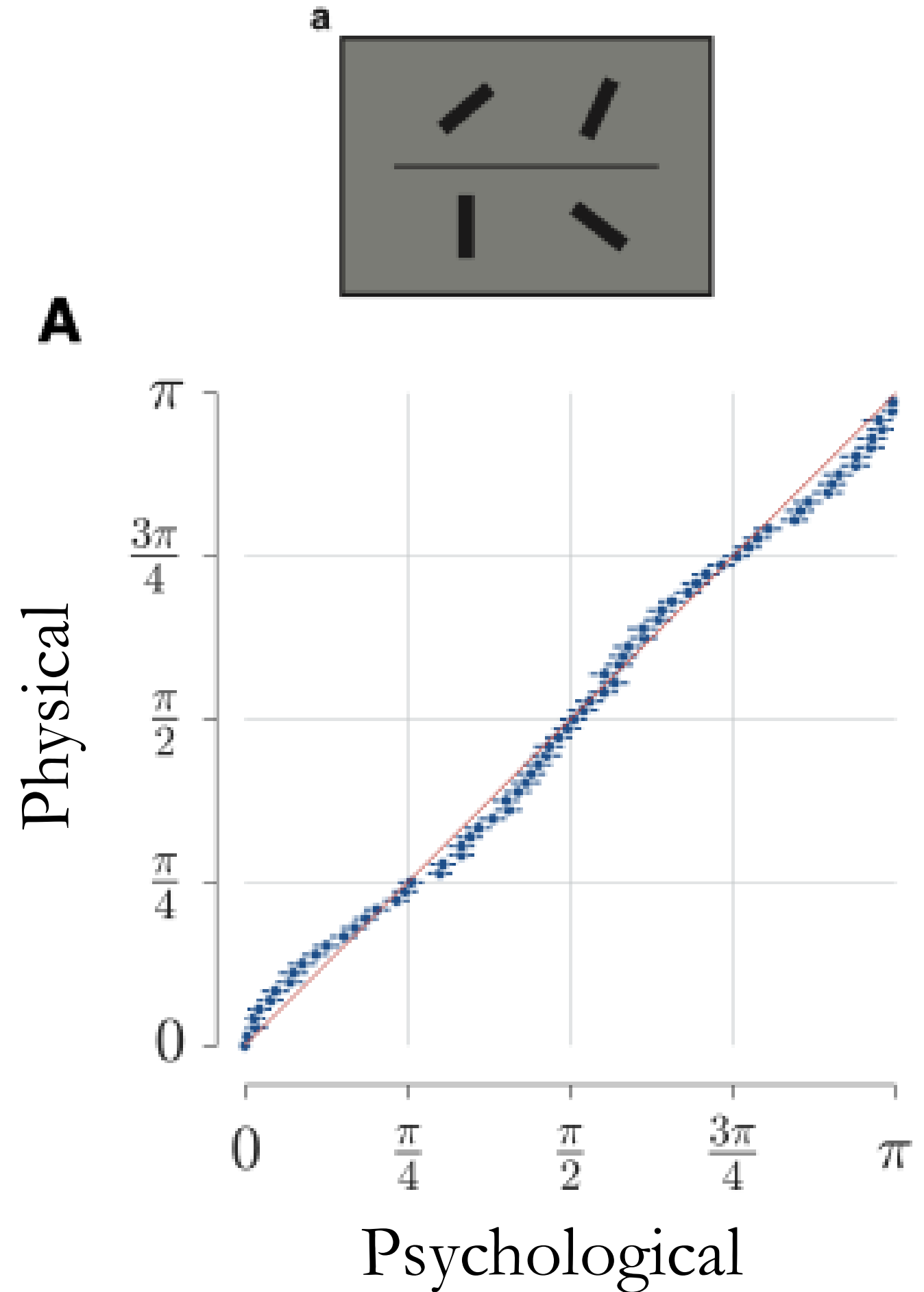


Memory reproduction

Similarity comparison

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

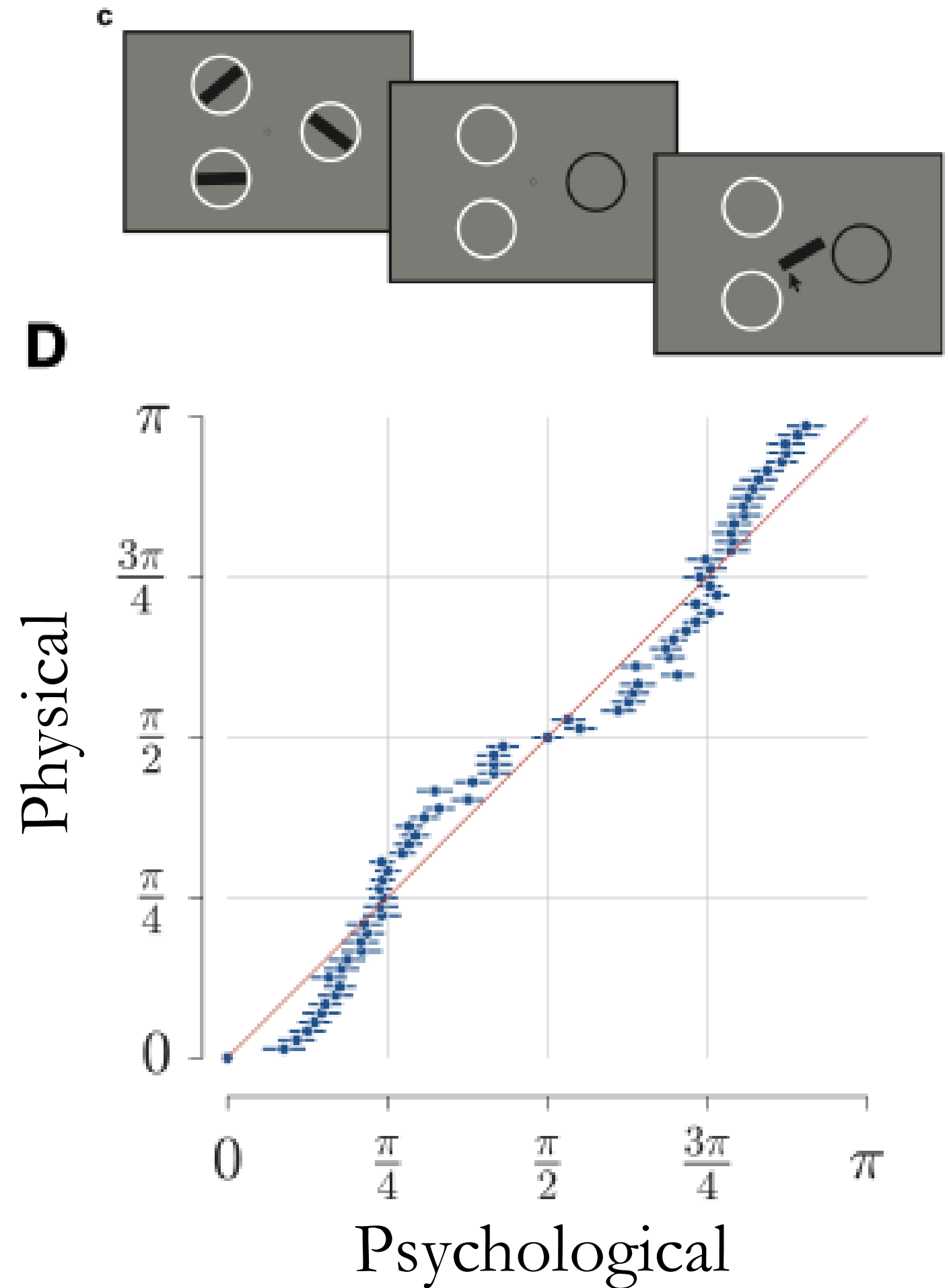
Clear deviations where close to vertical lines appear more vertical, and close to horizontal lines appear more horizontal.



Memory reproduction

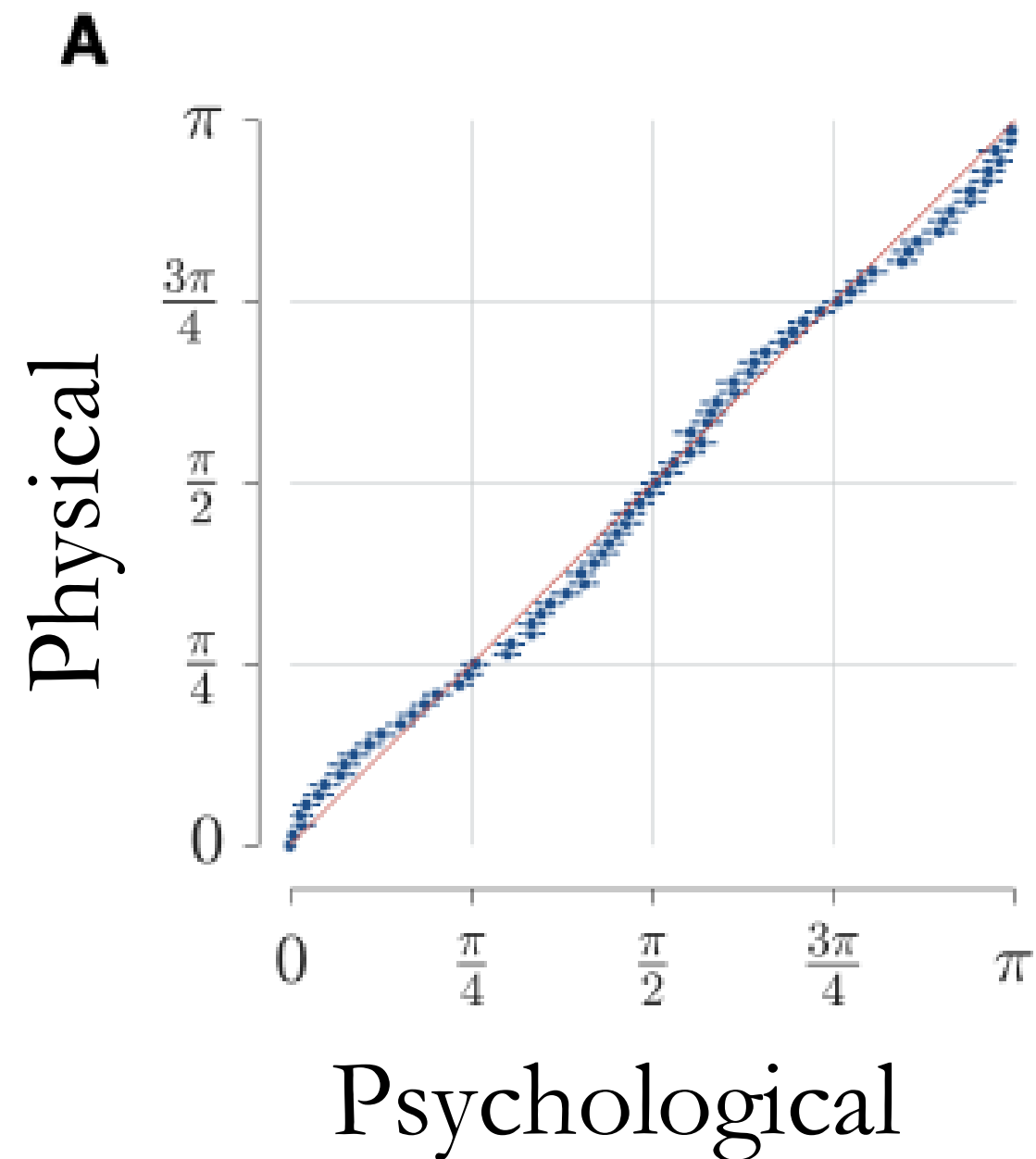
The representation for oriented lines do not exactly match the physically presented stimulus.

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

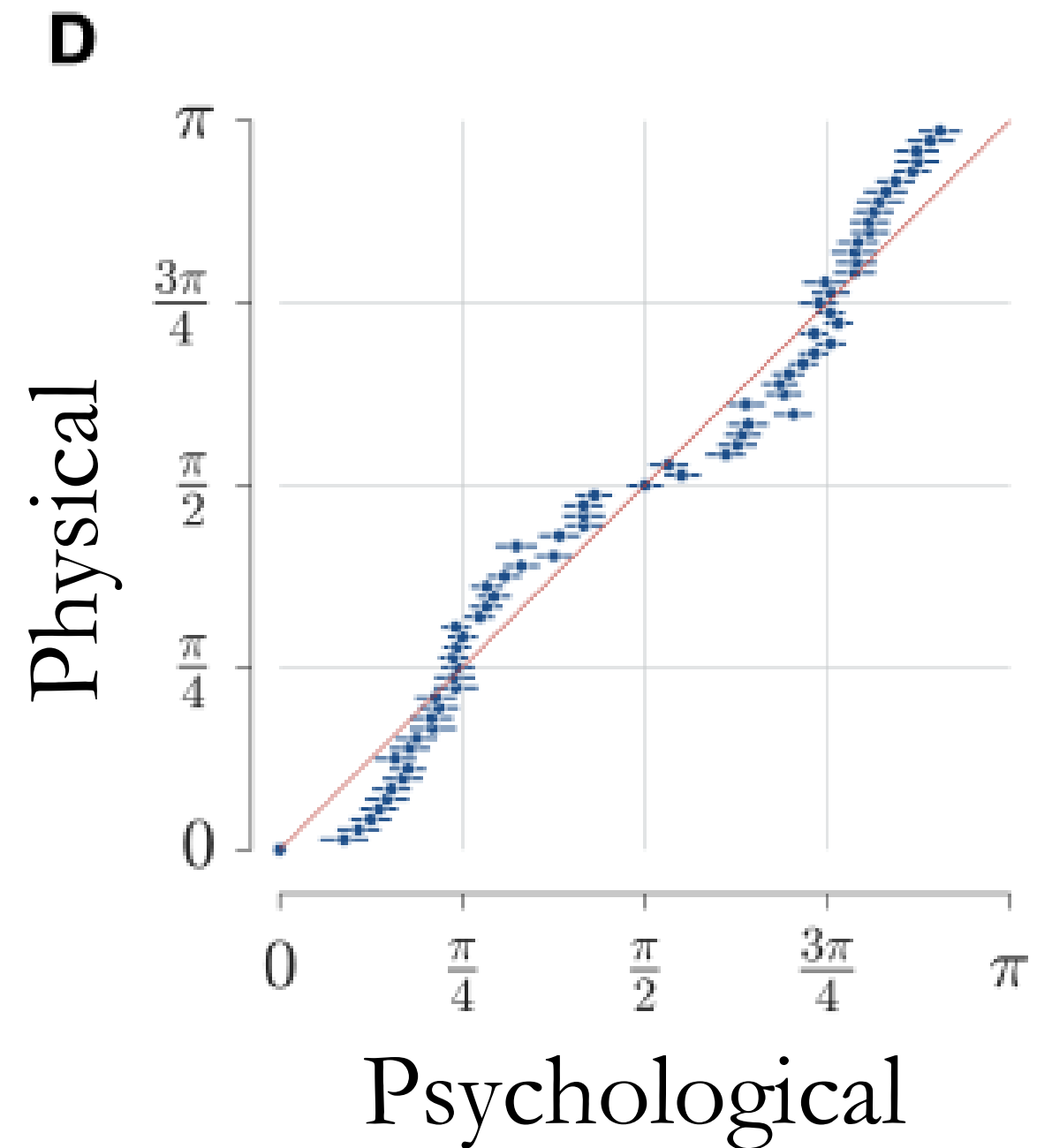


Similarity comparisons and both reproduction tasks **do not** share the same cognitive representation.

Similarity comparison



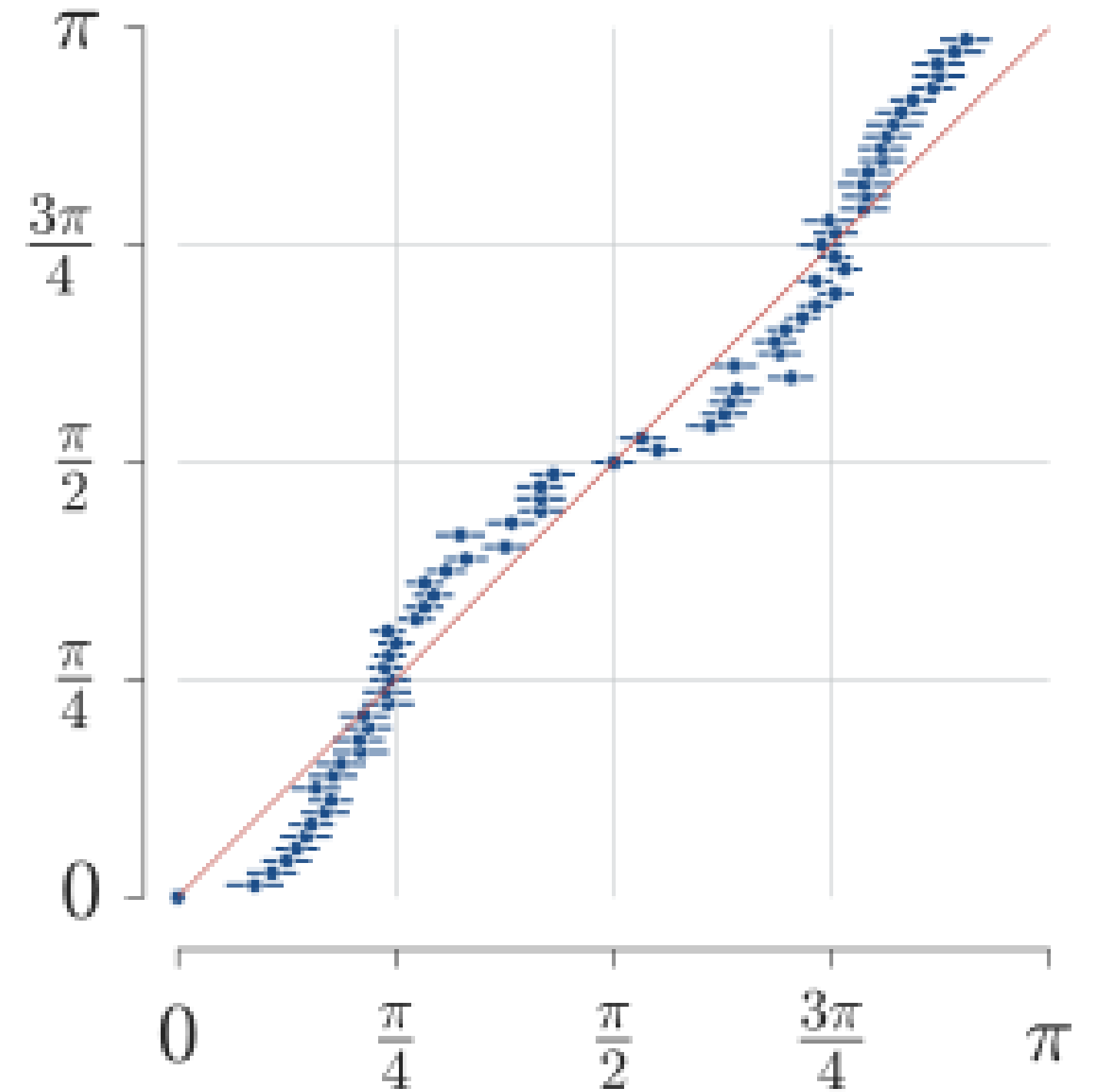
Memory reproduction



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



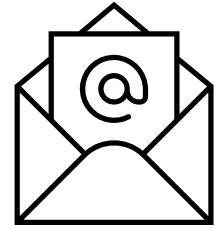
Takeaways

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!



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



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



Preprint

