

# Models of working memory aren't working for me

For the Visual Memory Lab, University of Nottingham  
21<sup>st</sup> March, 2025



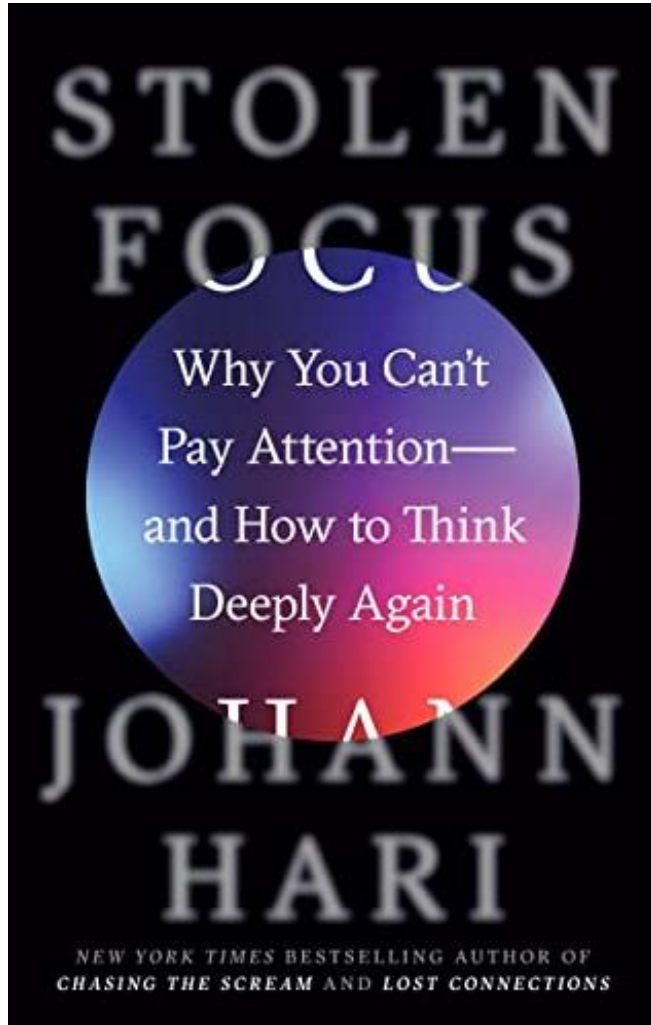
Why study attention and memory?

# Why study attention and memory?

- Our attention is very limited
- Therefore, our attention is *precious*
- We should want the right things to take up our attention!



# Why study attention and memory?

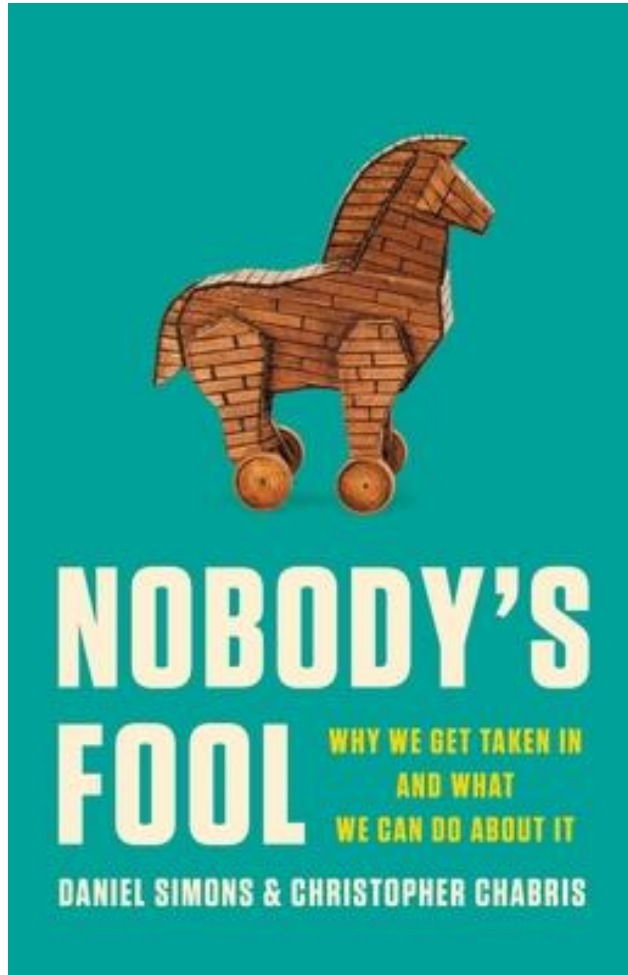


New York Times Bestseller,  
Book of the Year by Financial  
Times, etc.

Taps into the collective feeling that  
we are losing our ability to ***focus***

In my opinion, a very average book...

# Why study attention and memory?



We can be deceived when made to attend to the wrong things.

The researchers behind the  
“invisible gorilla” study!

What is working memory?

# What *is* visual working memory?

- “The system responsible for maintaining visual information in a state of heightened accessibility for ongoing perception and cognition.”
- This same definition could also describe visual **attention**
  - Perhaps also visual **imagery**, psychological **introspection**

# What *is* visual working memory?

- Many subtly different definitions:

## The many faces of working memory and short-term storage

[Nelson Cowan](#) 

[Psychonomic Bulletin & Review](#) **24**, 1158–1170 (2017) | [Cite this article](#)

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It has become clearer to me that a major source of confusion is that researchers use different definitions of the malleable and useful concept of WM. We do not seem to be converging on a common definition of the term. Others also have

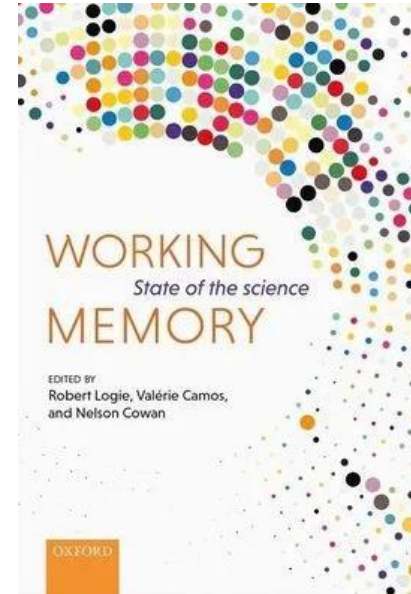


# What *is* visual working memory?

14

## Integrating Theories of Working Memory

*Robert H. Logie, Clément Belletier, and Jason M. Doherty*



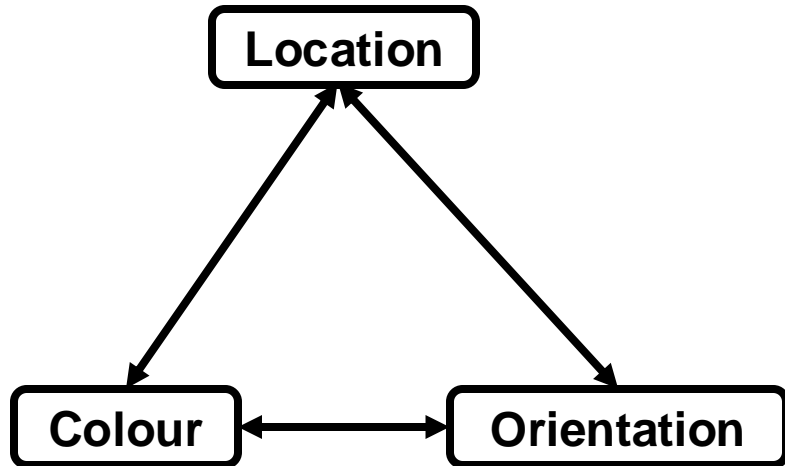
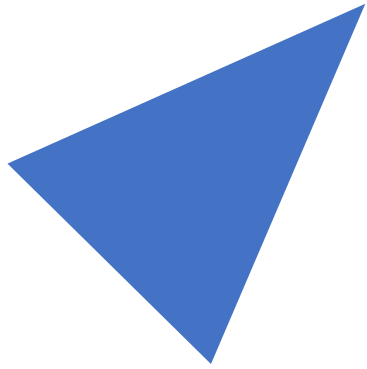
First published  
in late 2020

- “We argue that many of these differences reflect different research questions, different levels of explanation, differences in how participants perform their assigned tasks in different laboratories, **rather than fundamental theoretical adversity**”

# What *is* visual working memory?

- “The system responsible for maintaining visual information in a state of heightened accessibility for ongoing perception and cognition.”
- This same definition could also describe visual **attention**
  - Perhaps also visual **imagery**, psychological **introspection**
- The core question: **How is information represented in mind?**

# Representations in the mind



Constituents?  
Illusory objects?

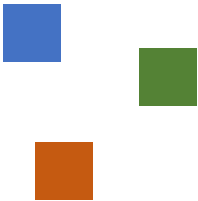
Memories across  
space and time?

# What *is* visual working memory?

## Object-based theory

*“slot models”*

(Luck and Vogel, 1997;  
Zhang and Luck, 2008)



## Feature-based theory

*“resource models”*

(Alvarez and Cavanagh, 2004;  
Wilken and Ma, 2004)



Luck, S. J., & Vogel, E. K. (1997). <https://doi.org/10.1038/36846>  
Zhang, W., & Luck, S. J. (2008). <https://doi.org/10.1038/nature06860>  
Alvarez, G. A., & Cavanagh, P. (2004). <https://doi.org/10.1111/j.0963-7214.2004.01502006.x>  
Wilken, P., & Ma, W. J. (2004). <https://doi.org/10.1167/4.12.11>

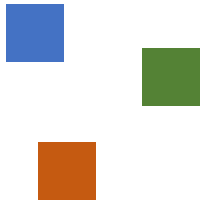
# What *is* visual working memory?

- An enduring theoretical framework has been

## Object-based theory

*“slot models”*

(Luck and Vogel, 1997;  
Zhang and Luck, 2008)



versus

## Feature-based theory

*“resource models”*

(Alvarez and Cavanagh, 2004;  
Wilken and Ma, 2004)



# *A theory crisis* in psychological science

- An understated precursor to the *reproducibility crisis* may be the lack of coordinated theoretical development
  - An over-reliance on the hypothetico-deductive method (e.g. null hypothesis significance testing) for inferences
    - Questionable research practices (QRPs): *p*-hacking, HARKing, data manipulation, etc.

Borsboom D. (2013, November 20). Theoretical amnesia. *Center for Open Science*

Borsboom, D., van der Maas, H. L., Dalege, J., Kievit, R. A., & Haig, B. D. (2021). Theory construction methodology: A practical framework for building theories in psychology. *Perspectives on Psychological Science*, 16(4), 756-766.

Oberauer K., Lewandowsky S. (2019). Addressing the theory crisis in psychology. *Psychonomic Bulletin & Review*, 26, 1596–1618.

Maatman, F. O. (2021). Psychology's theory crisis, and why formal modelling cannot solve it. *PsyArXiv*

Meehl P. E. (1978). Theoretical risks and tabular asterisks: Sir Karl, Sir Ronald, and the slow progress of soft psychology. *Journal of Consulting and Clinical Psychology*, 46, 806–834.

# Playing *20 questions* with nature

- It is often assumed that...

Theory A



Result A

Theory B



Result B

# Playing *20 questions* with nature

- It is often assumed that...

Theory A



~~Result A~~

 Theory B

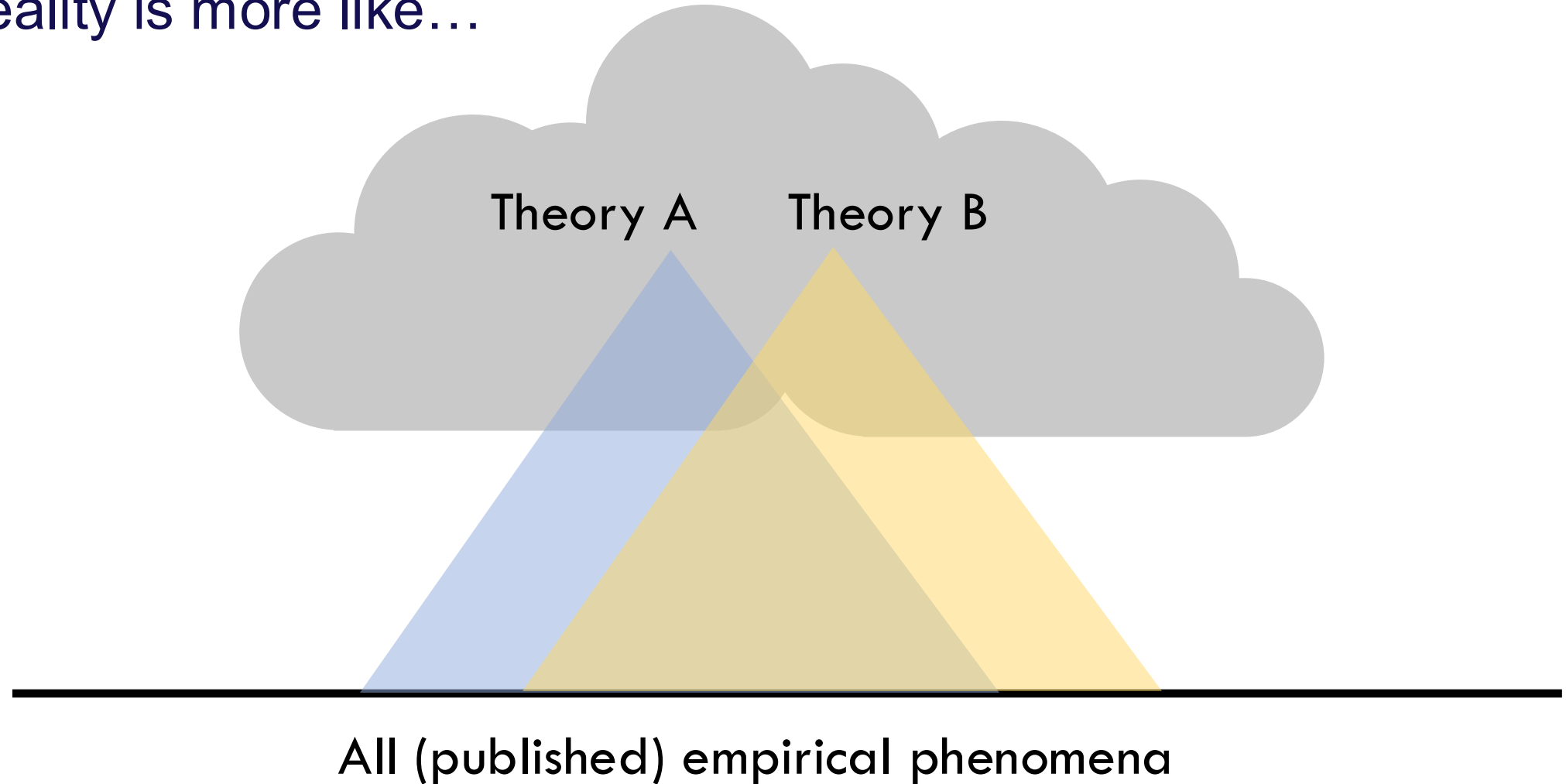


Result B



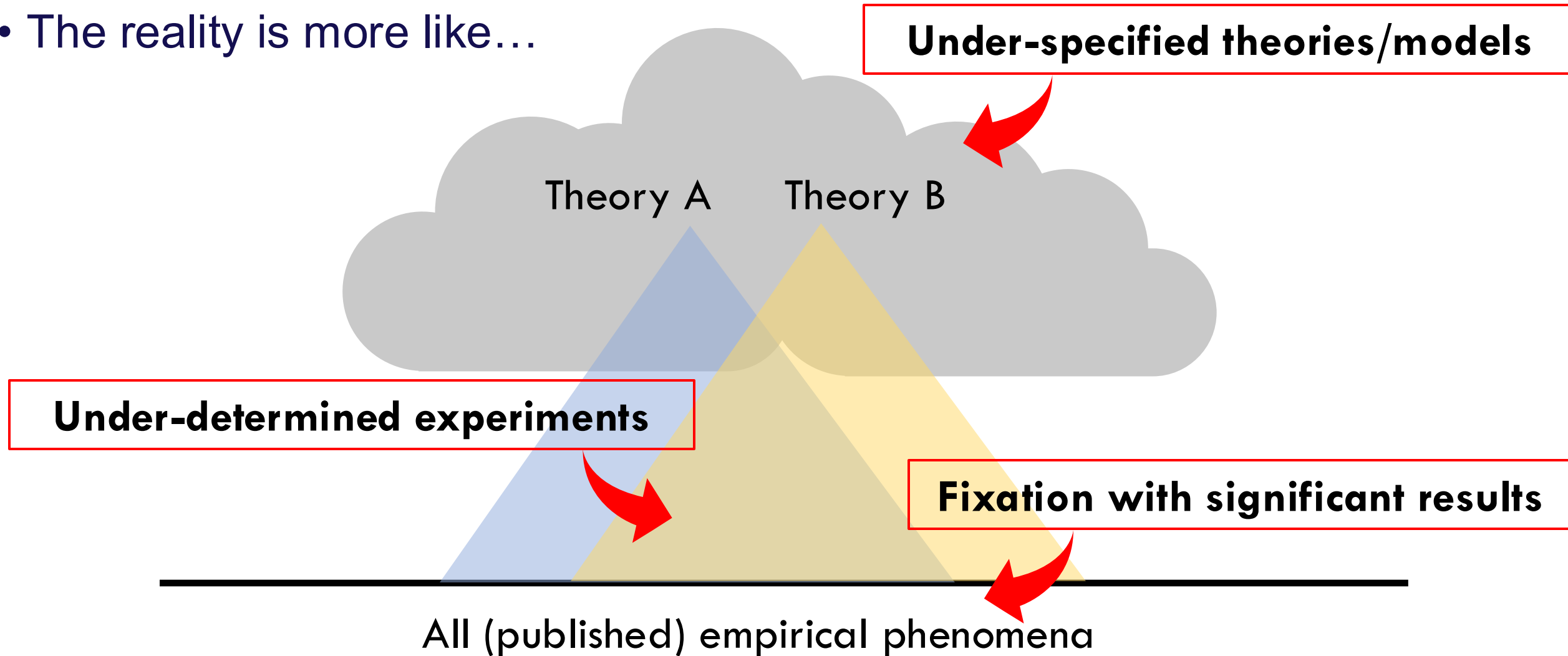
# Playing *20 questions* with nature

- The reality is more like...



# Playing *20 questions* with nature

- The reality is more like...



# *A theory crisis* in psychological science

- An understated precursor to the *reproducibility crisis* may be the lack of coordinated theoretical development
  - An over-reliance on the hypothetico-deductive method (e.g. null hypothesis significance testing) for inferences
    - Questionable research practices (QRPs): *p*-hacking, HARKing, data manipulation, etc.
  - Under-specified theories with under-determined experimental designs
    - *Ad hoc* changes in models, straw-man of competing models, blunt instruments of measurement
  - Overgeneralization of a theory or model to all related phenomena or empirical conditions
    - A lack of intellectual humility...

Borsboom D. (2013, November 20). Theoretical amnesia. *Center for Open Science*

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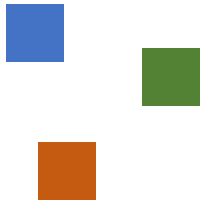
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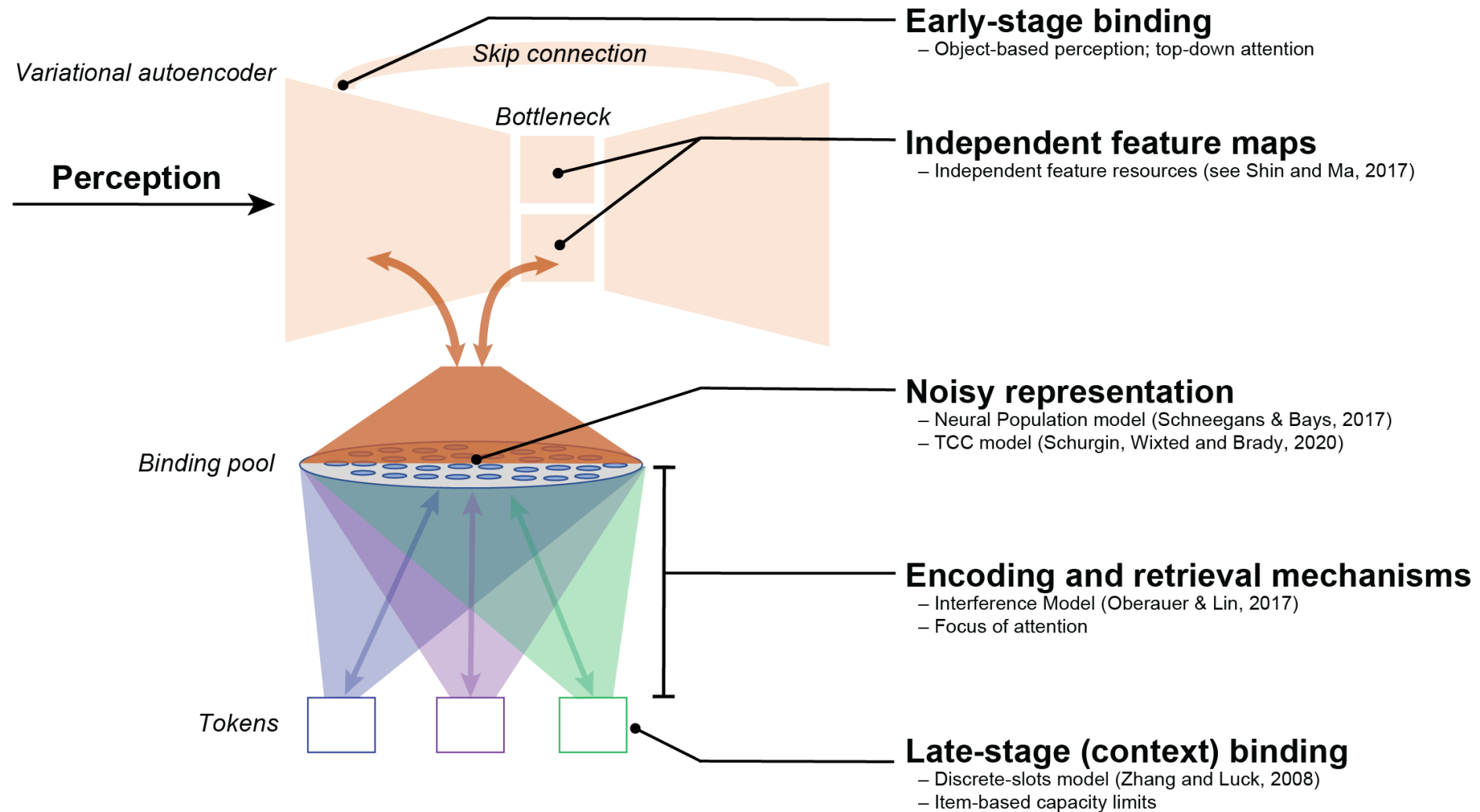
How do we address the theory crisis?

Can we bring these models into accordance?

# Towards a model-centric science

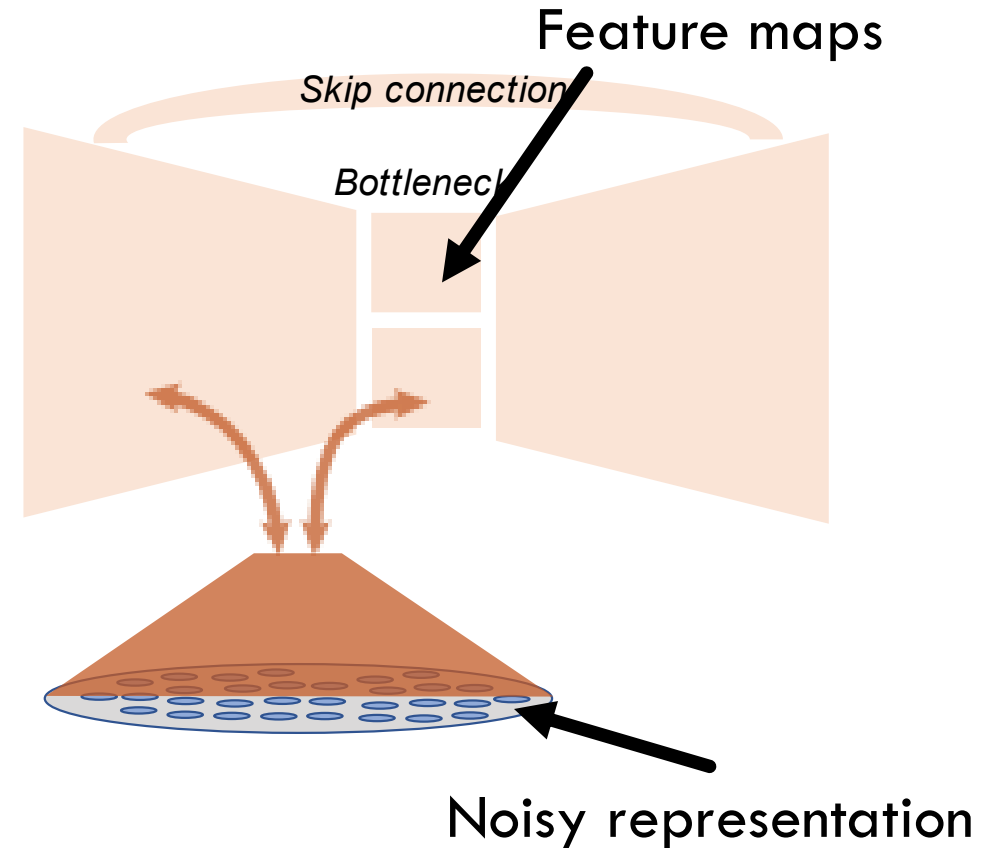
- We need to move away from dualistic experiments and a results-oriented science towards a model-centric science
- We need more theory development
  - Repeating and detailing the phenomena that we hope to explain
  - Integrating various empirical results and models
  - Clear specification of theories and models and how they relate to the phenomena
  - Careful generalization of current models (i.e. practicing intellectual humility)
  - Better thought-out methods and measures
  - Rigorous design of experiments to truly test hypotheses

# Presenting a **theory map** for visual working memory



# Binding pool as a locus for feature-based ideas

- Independent feature layers project into the binding pool (Shin and Ma, 2017)
  - But early-stage object-based attention may also be in play
- **Noisy representations** in VWM are well-captured by neural population and signal detection accounts (Bays, 2014; Schurgin et al., 2020)



Bays, P. M. (2014). Noise in neural populations accounts for errors in working memory. *Journal of Neuroscience*, 34(10), 3632-3645.

Schurgin, M. W., Wixted, J. T., & Brady, T. F. (2020). Psychophysical scaling reveals a unified theory of visual memory strength. *Nature human behaviour*, 4(11), 1156-1172.

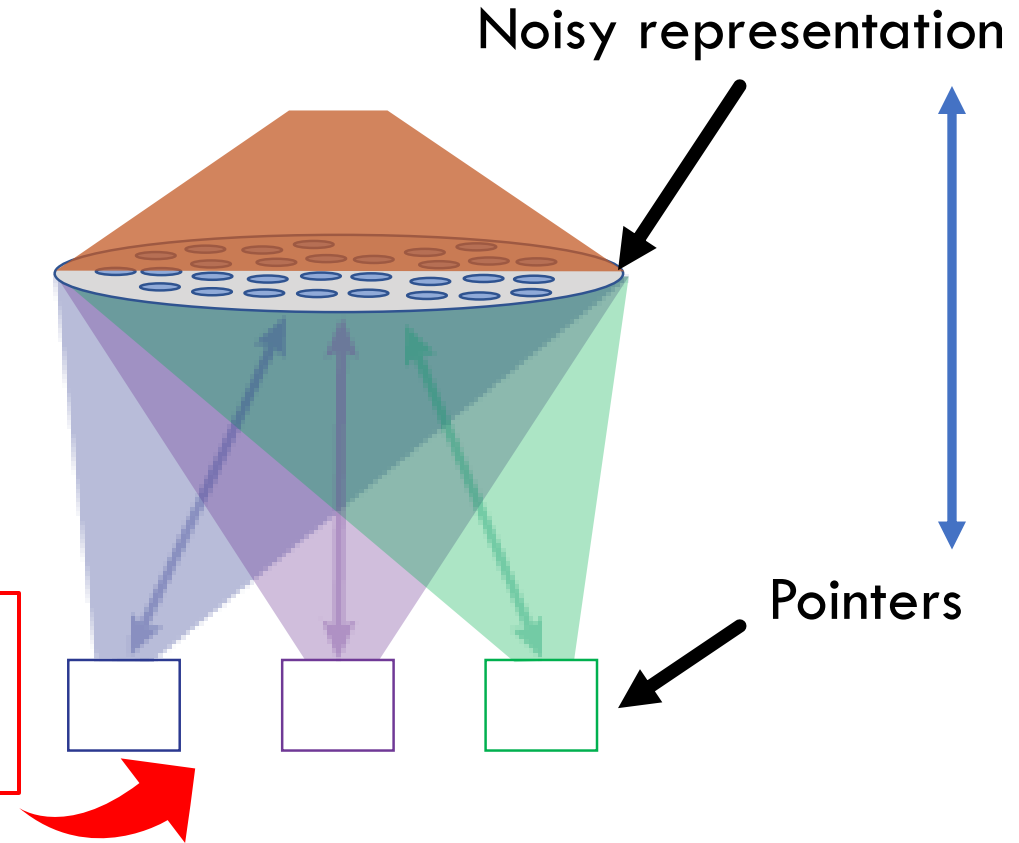
Shin, H., & Ma, W. J. (2017). Visual short-term memory for oriented, colored objects. *Journal of Vision*, 17(9), 12-12.



# Tokens as a locus for object-based ideas

- **Content-independent pointers**
  - Like *FINSTs* or *Object Files* (Pylyshyn, 1989; Kahneman et al., 1992)
- Evidence for a neural signature that indexes VWM load and generalizes across feature content (Thyer et al., 2022; Balaban et al., 2019)

**New conception of working memory as a very late-stage of encoding and selection**



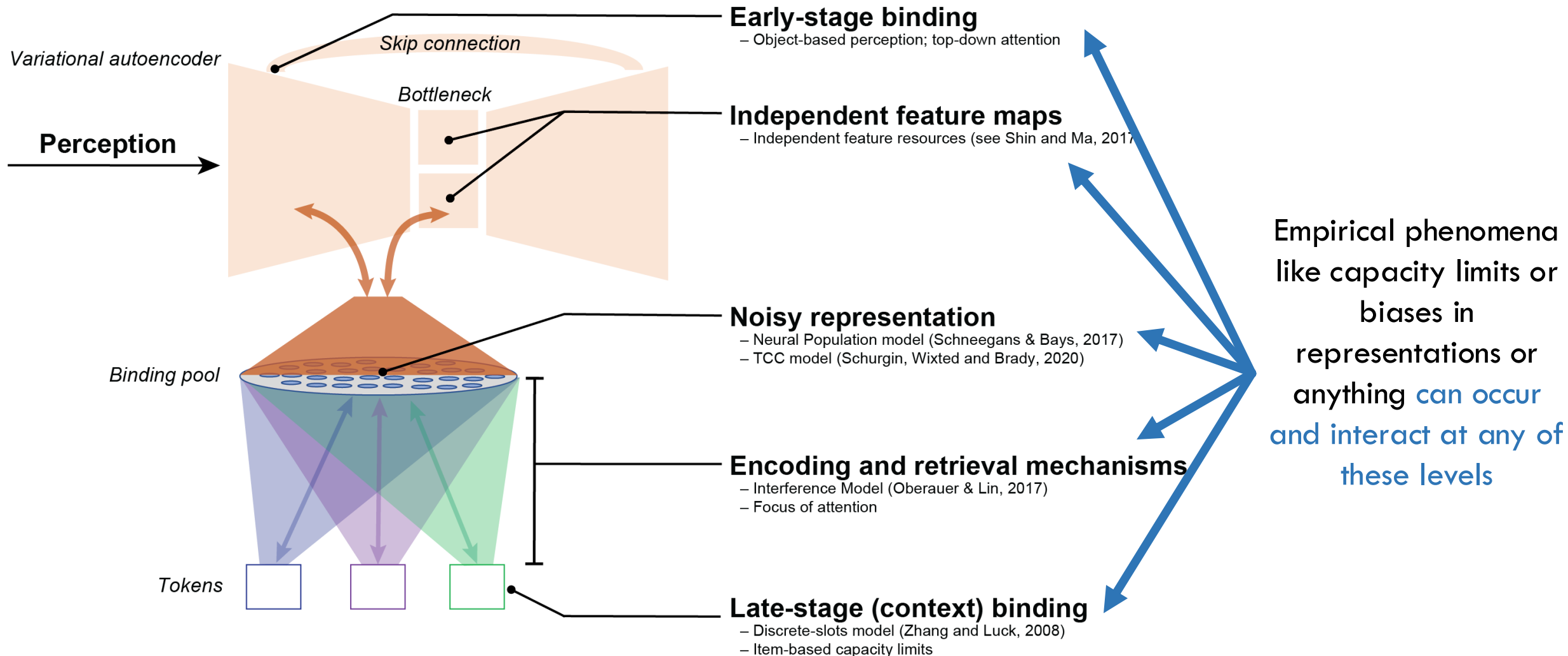
Pylyshyn, Z. (1989). The role of location indexes in spatial perception: A sketch of the FINST spatial-index model. *Cognition*, 32(1), 65-97.

Kahneman, D., Treisman, A., & Gibbs, B. J. (1992). The reviewing of object files: Object-specific integration of information. *Cognitive psychology*, 24(2), 175-219.

Thyer, W., Adam, K. C., Diaz, G. K., Velazquez Sanchez, I. N., Vogel, E. K., & Awh, E. (2022). Storage in visual working memory recruits a content-independent pointer system. *Psychological Science*, 33(10), 1680-1694.

Balaban, H., Drew, T., & Luria, R. (2019). Neural evidence for an object-based pointer system underlying working memory. *cortex*, 119, 362-372.

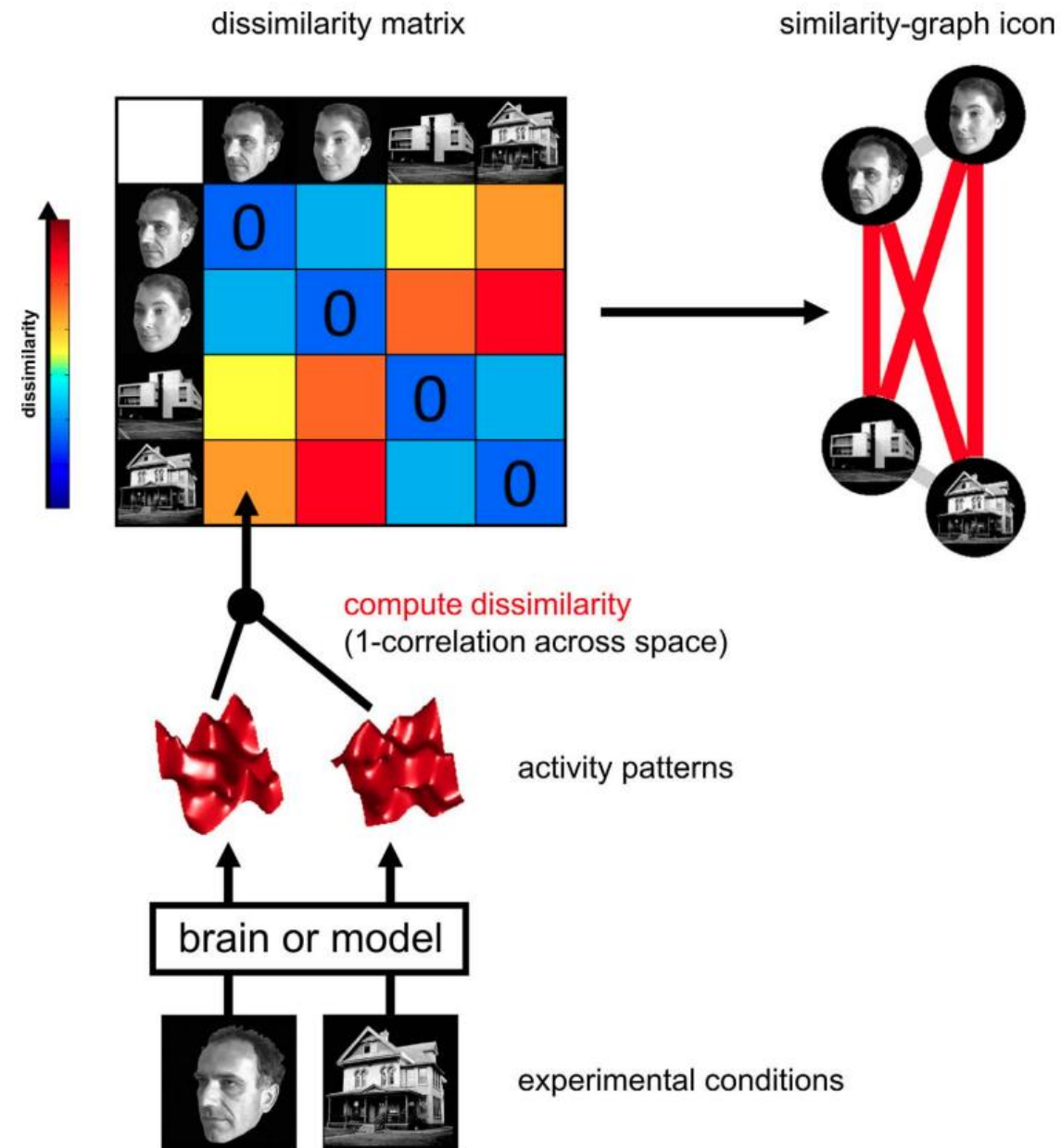
# Presenting a **theory map** for visual working memory



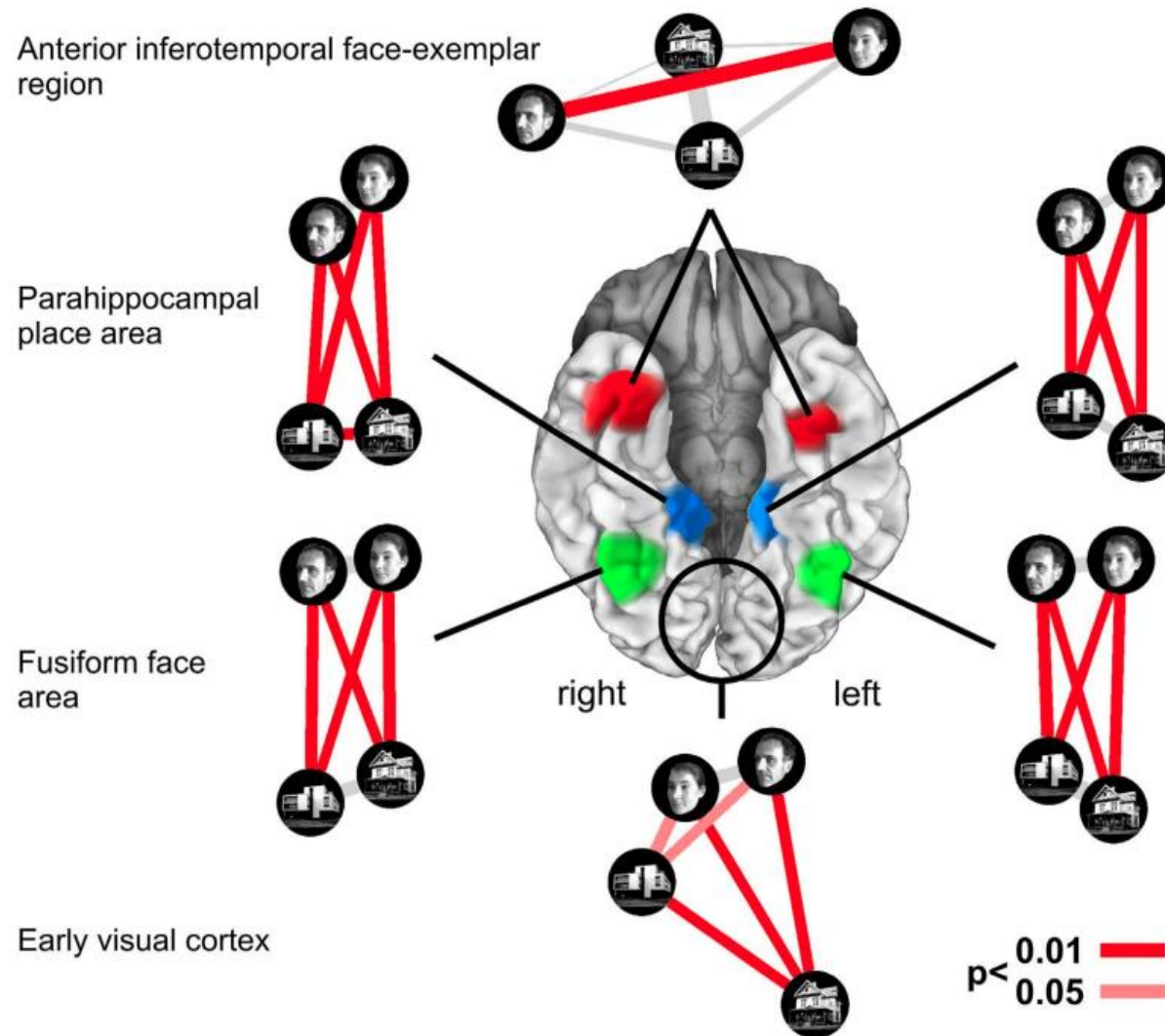
# How does a theory map help?

- Provides a **common core language and framework** to discuss theories, models, and phenomena
  - Reveals hidden intuitions
  - Prevents misunderstandings from varying definitions
  - Better specifies connection between models and phenomena
  - Reduces straw-man of various positions
  - Discourages a dualistic framework for experimental design
  - Initiates better determined model comparisons and definitive empirical tests
- Inspires **theory development**
  - Promotes **counterinduction** (the use and development of others' models)
  - Encourages **slow science** from better thought-out studies

# A brief overview of multivariate decoding

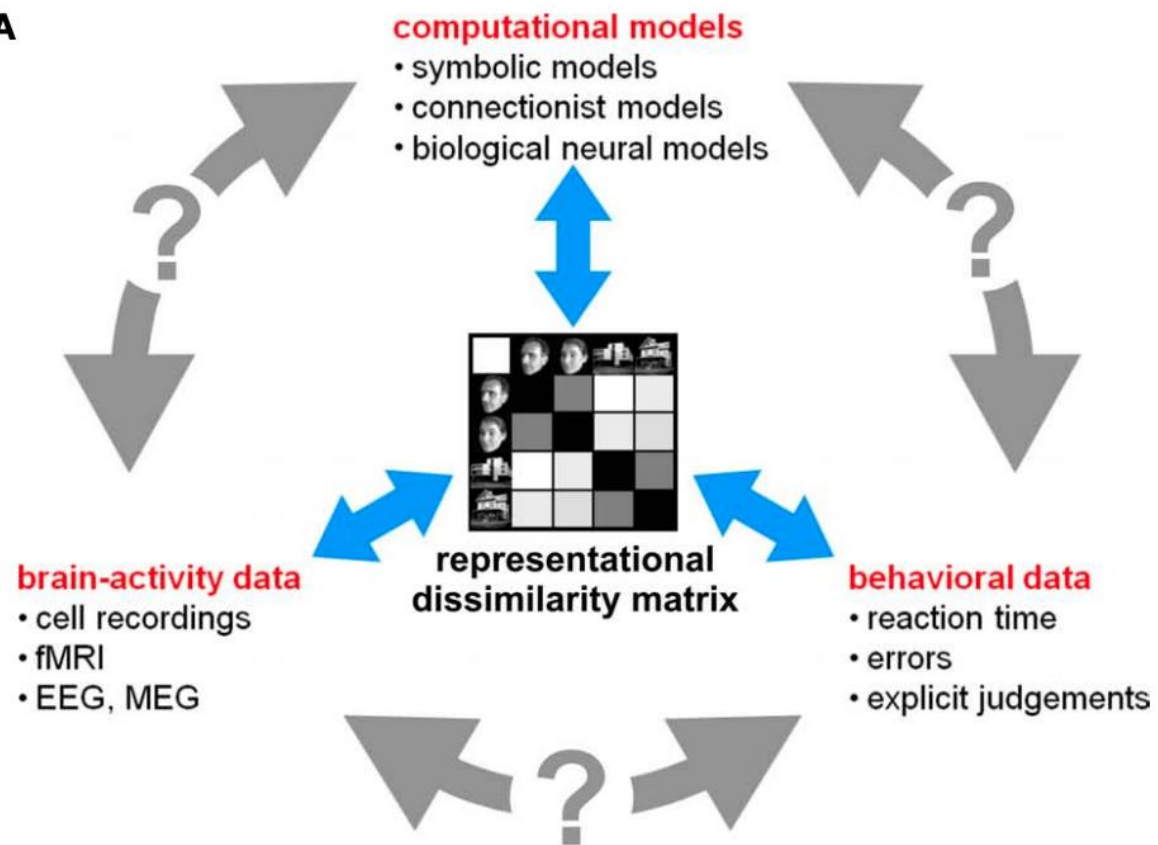


Kriegeskorte, N., Mur, M., & Bandettini, P. A. (2008). Representational similarity analysis-connecting the branches of systems neuroscience. *Frontiers in systems neuroscience*, 2, 249.

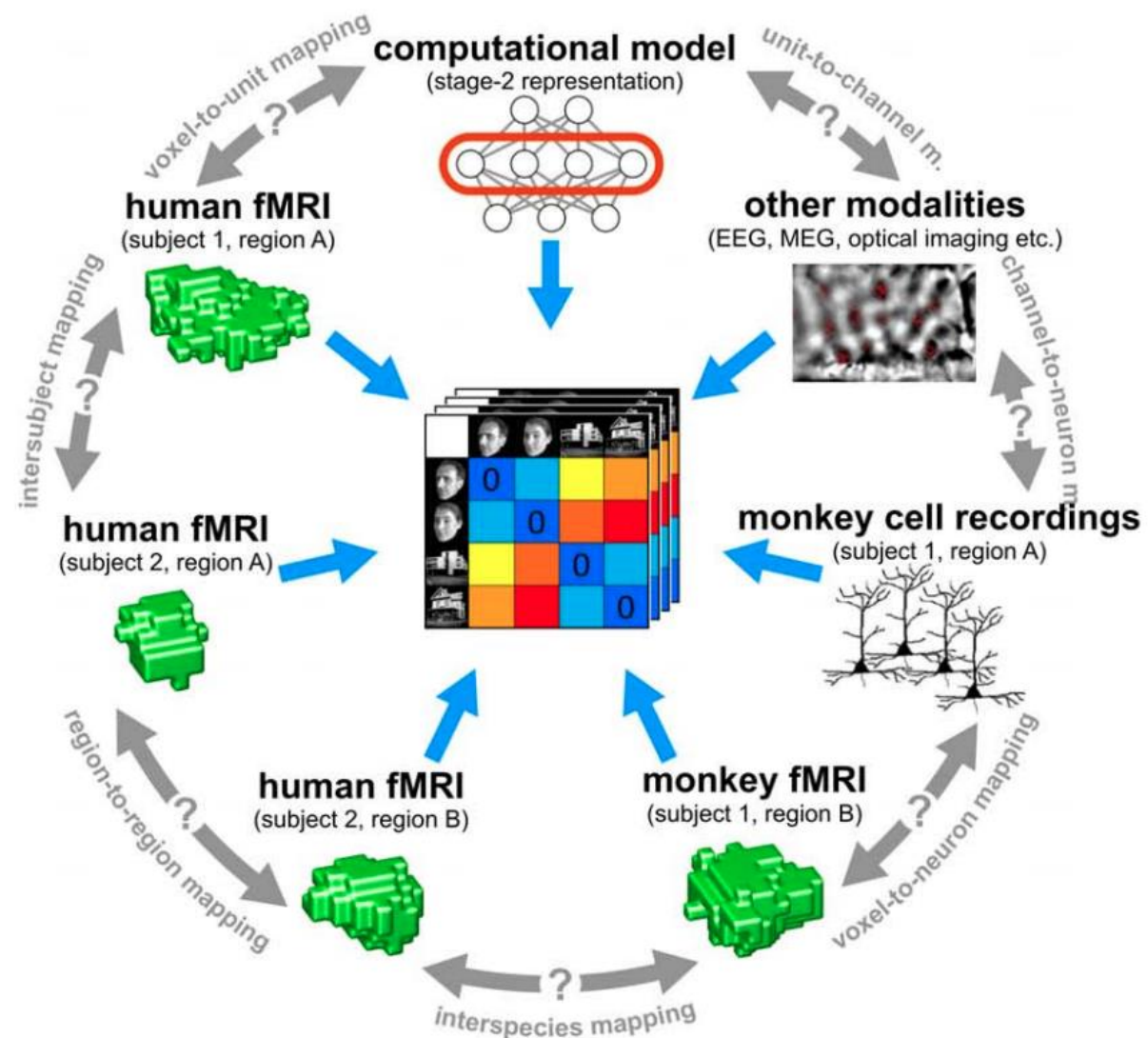


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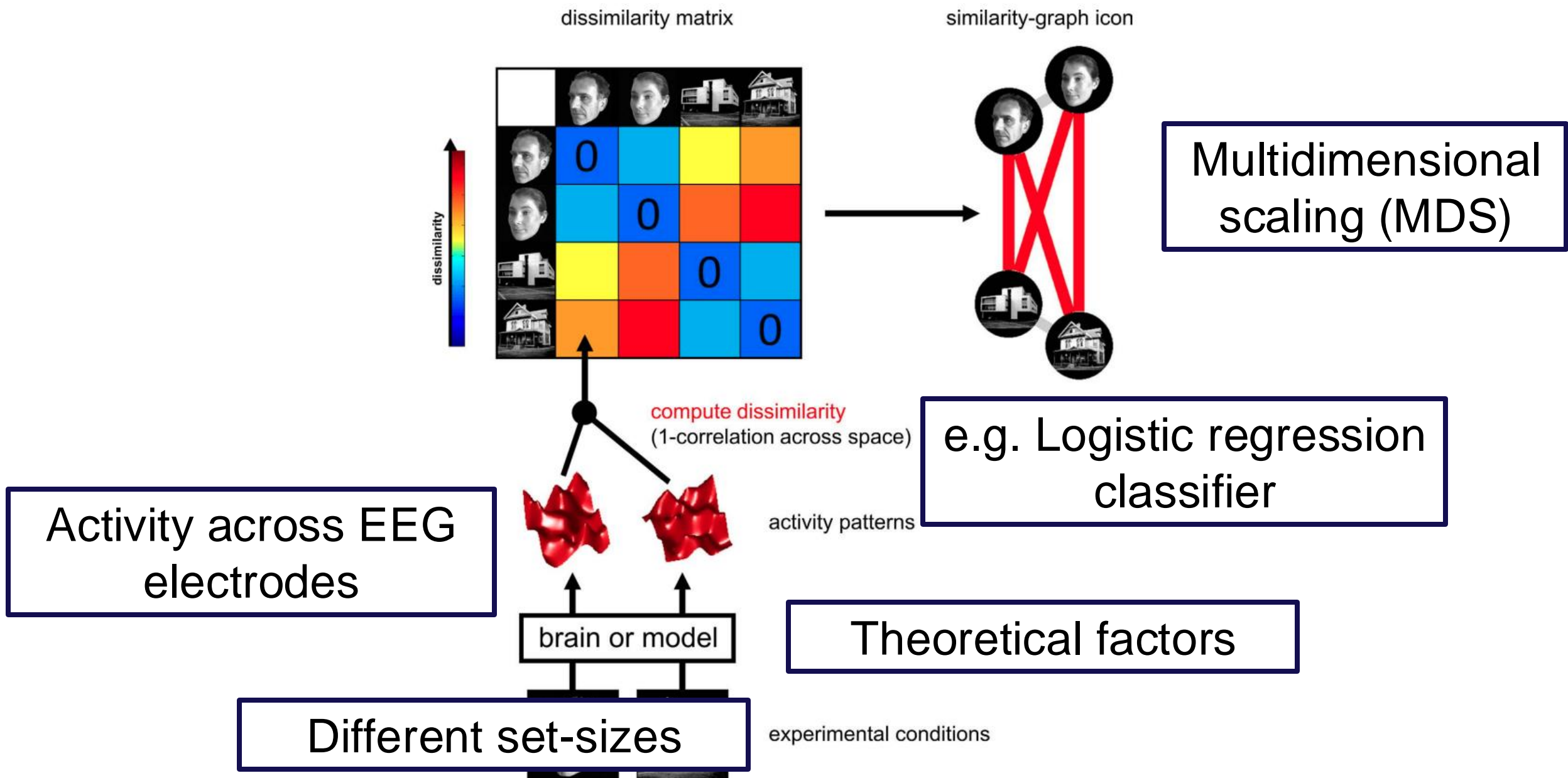
**A**



**B**





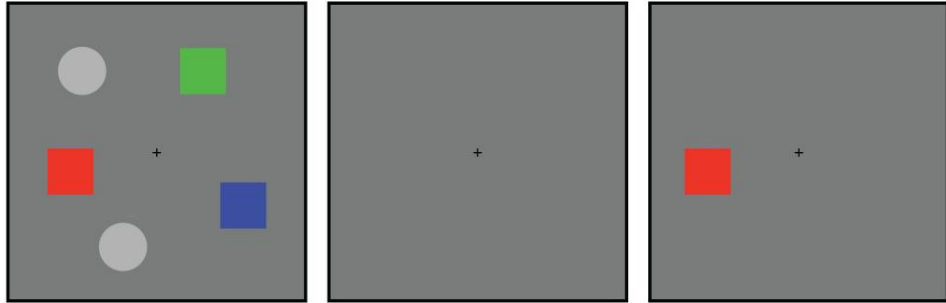


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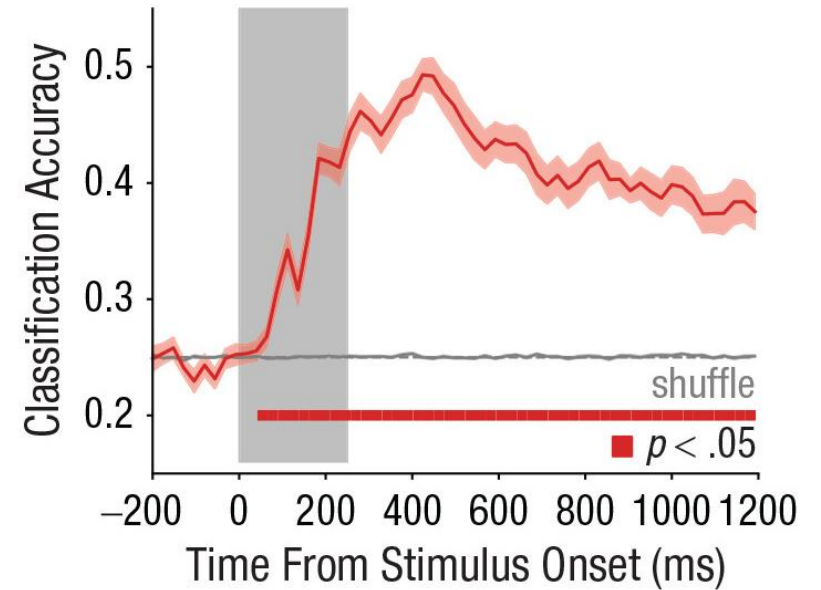


# Multivariate classification of working memory

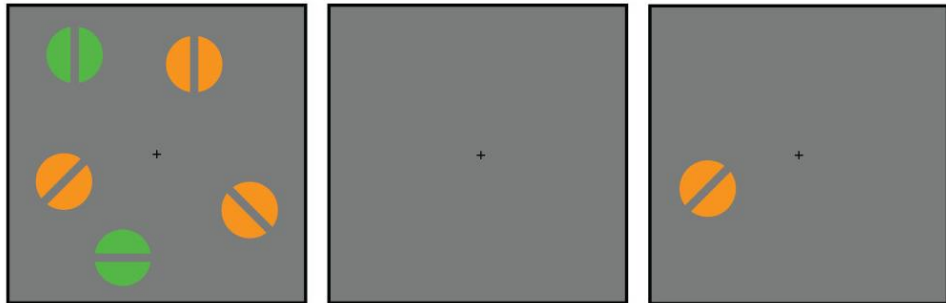
Experiment 1: Color



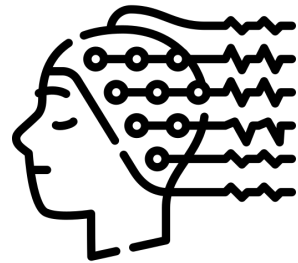
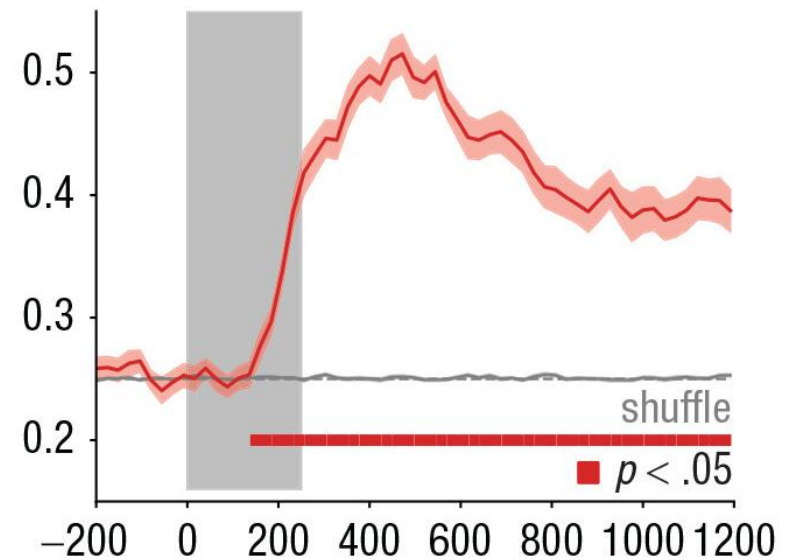
Train and test



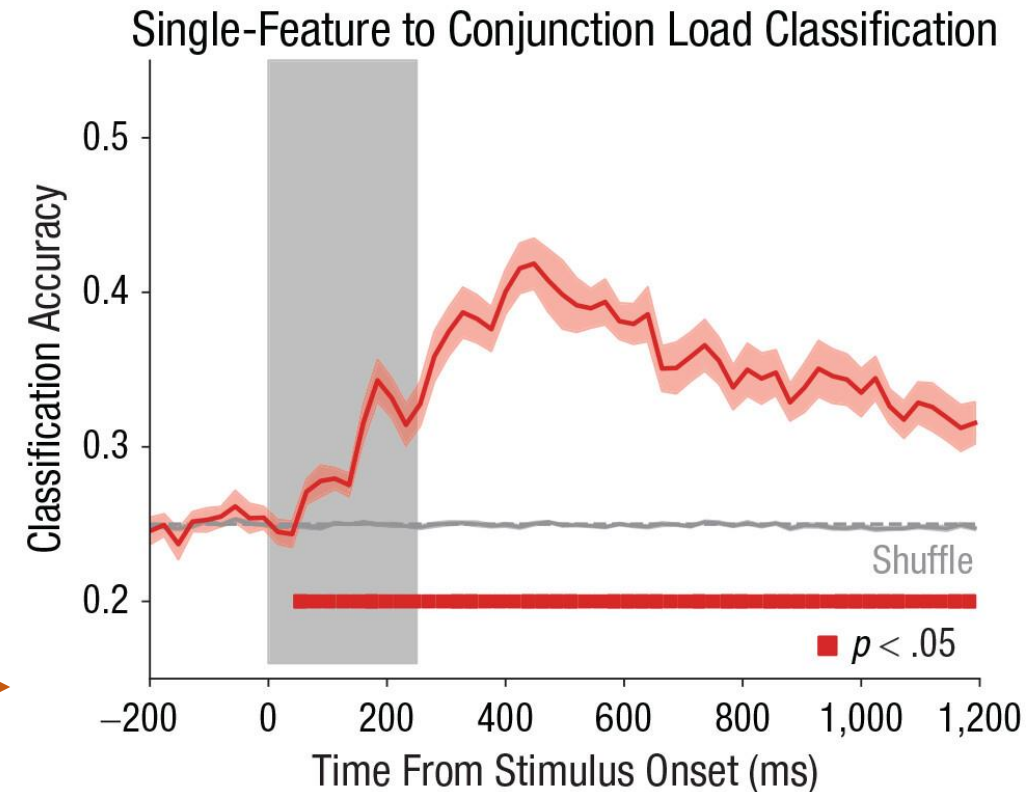
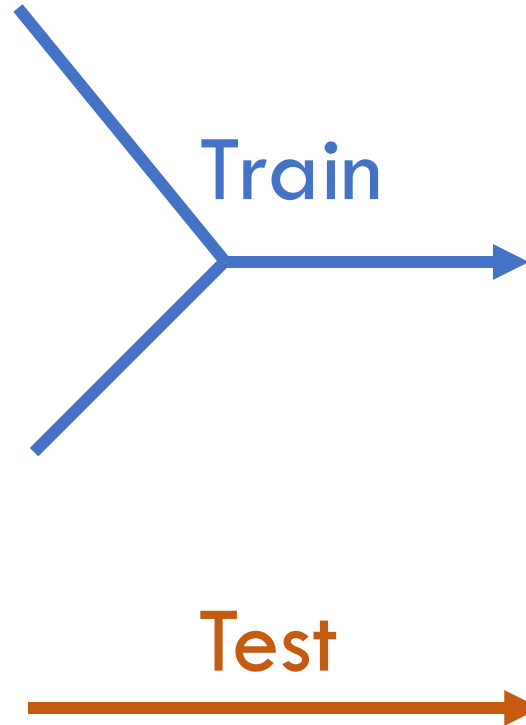
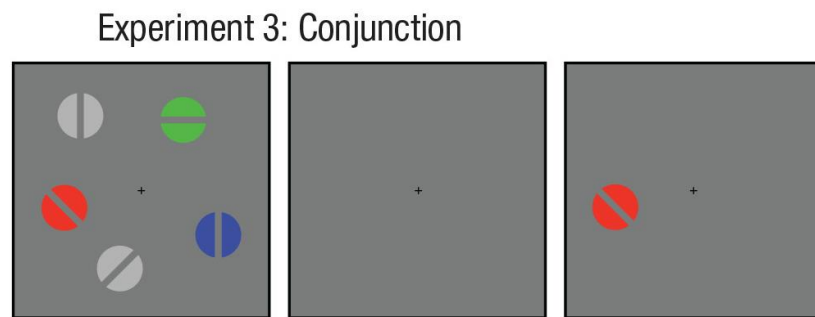
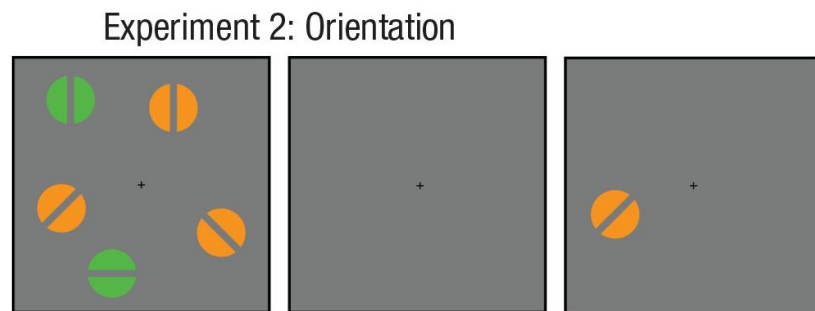
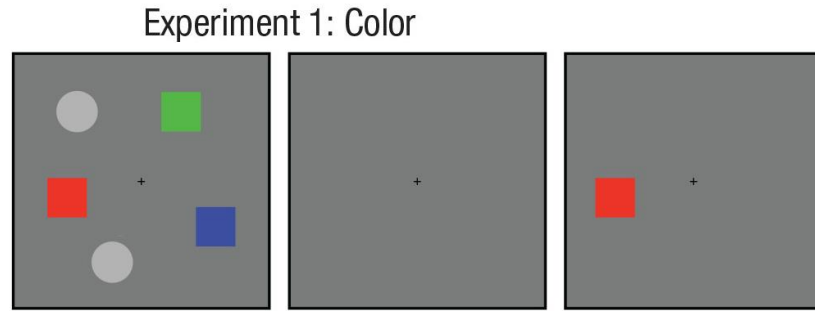
Experiment 2: Orientation



Train and test

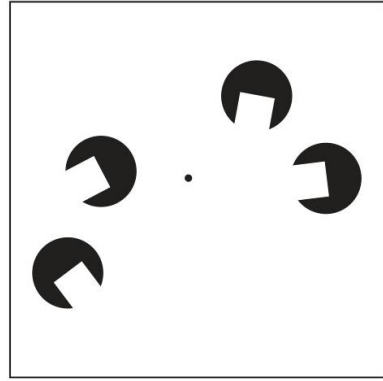


# Multivariate classification of working memory

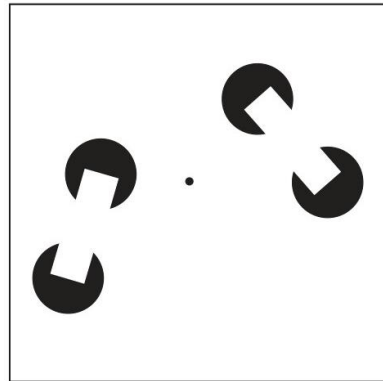


# Multivariate classification of working memory

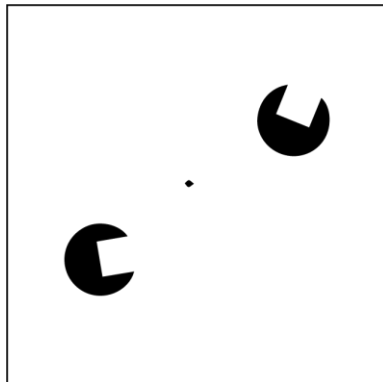
**4 Ungrouped**



**4 Grouped**



**2 Ungrouped**

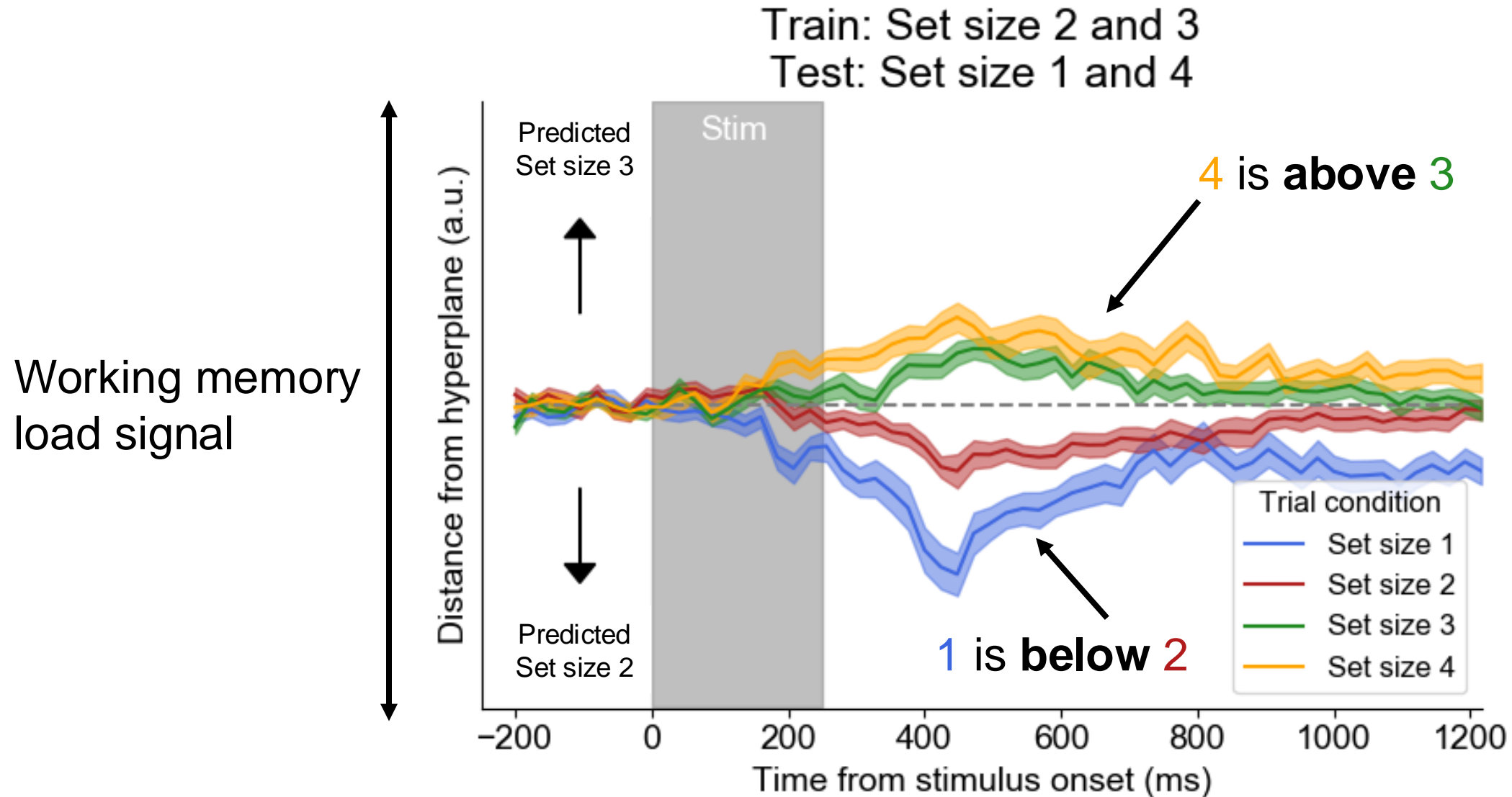


Distance from hyperplane (a.u.)

-250 0 250 500 750 1000 1250 1500  
Time from stimulus onset (ms)

Thyer et al. (2022) *Psychological Science*

# Multivariate classification of working memory



# Does learning change working memory?



Will  
Epstein



Henry  
Jones



Darius  
Suplica



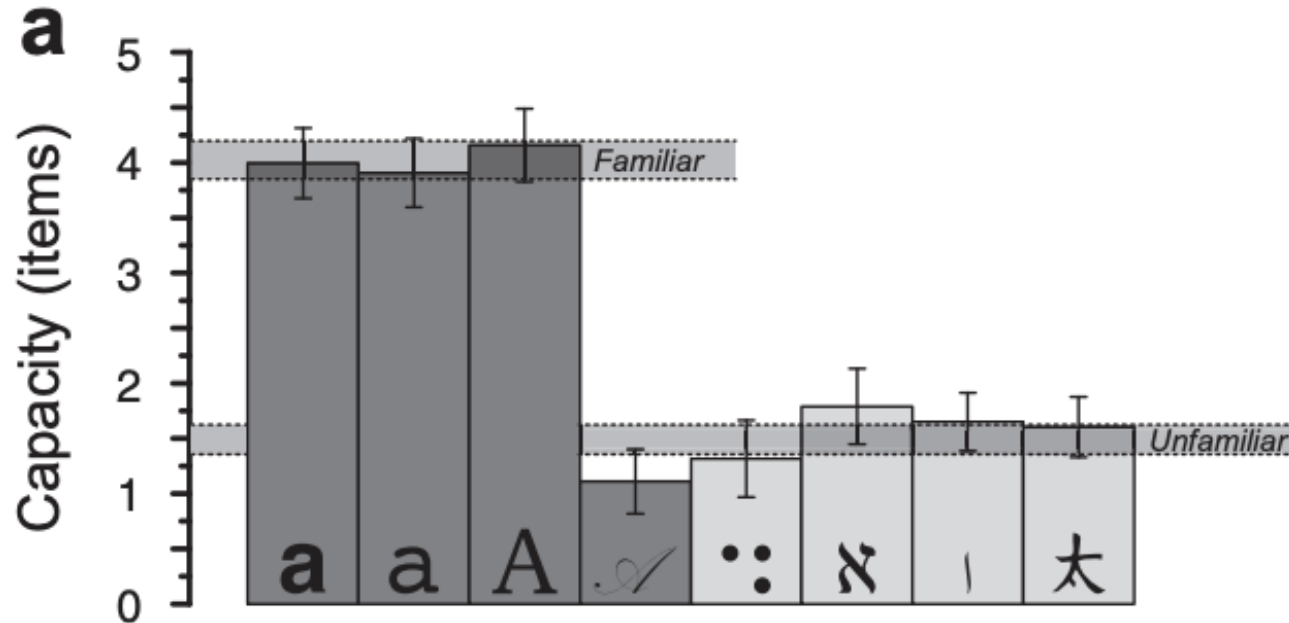
William  
Thyer



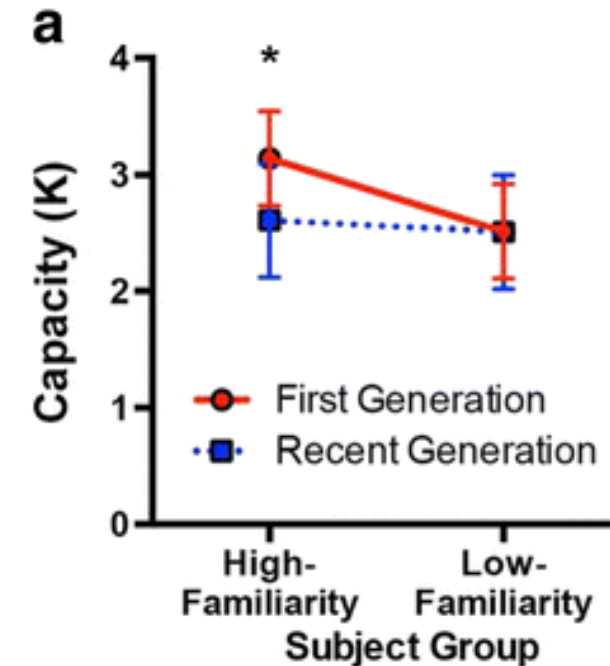
Edward  
Awh

# Working memory is aided by long-term memory

- A hallmark of our visual working memory system is its sharp capacity limit
- But this capacity limit can be overcome with **familiarity**:



Ngiam et al. (2019) *JEP:G*

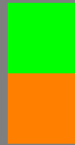


Xie and Zhang (2017) *M&C*



# Experiment 1: Training

- Subjects completed 600 trials to learn four color pairs:



# Experiment 1: Training





# Experiment 1: Training

- Two alternative-forced choice – which color was in the **bolded** location?



# Experiment 1: Pre-training and post-training

- Before training – 4 random colors
- After training – 4 paired colors (two learned pairs)



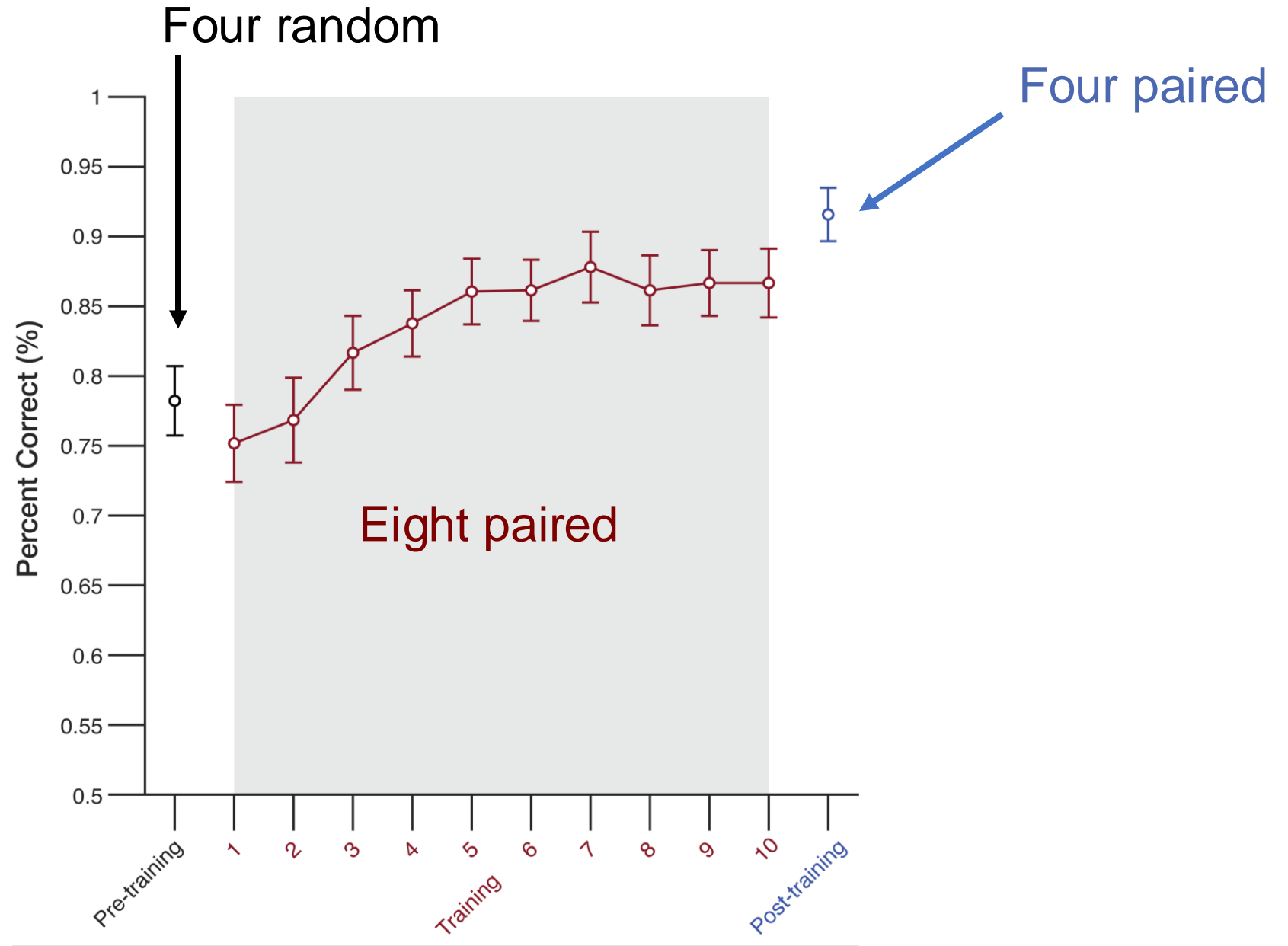
# Experiment 1: Pre-training and post-training



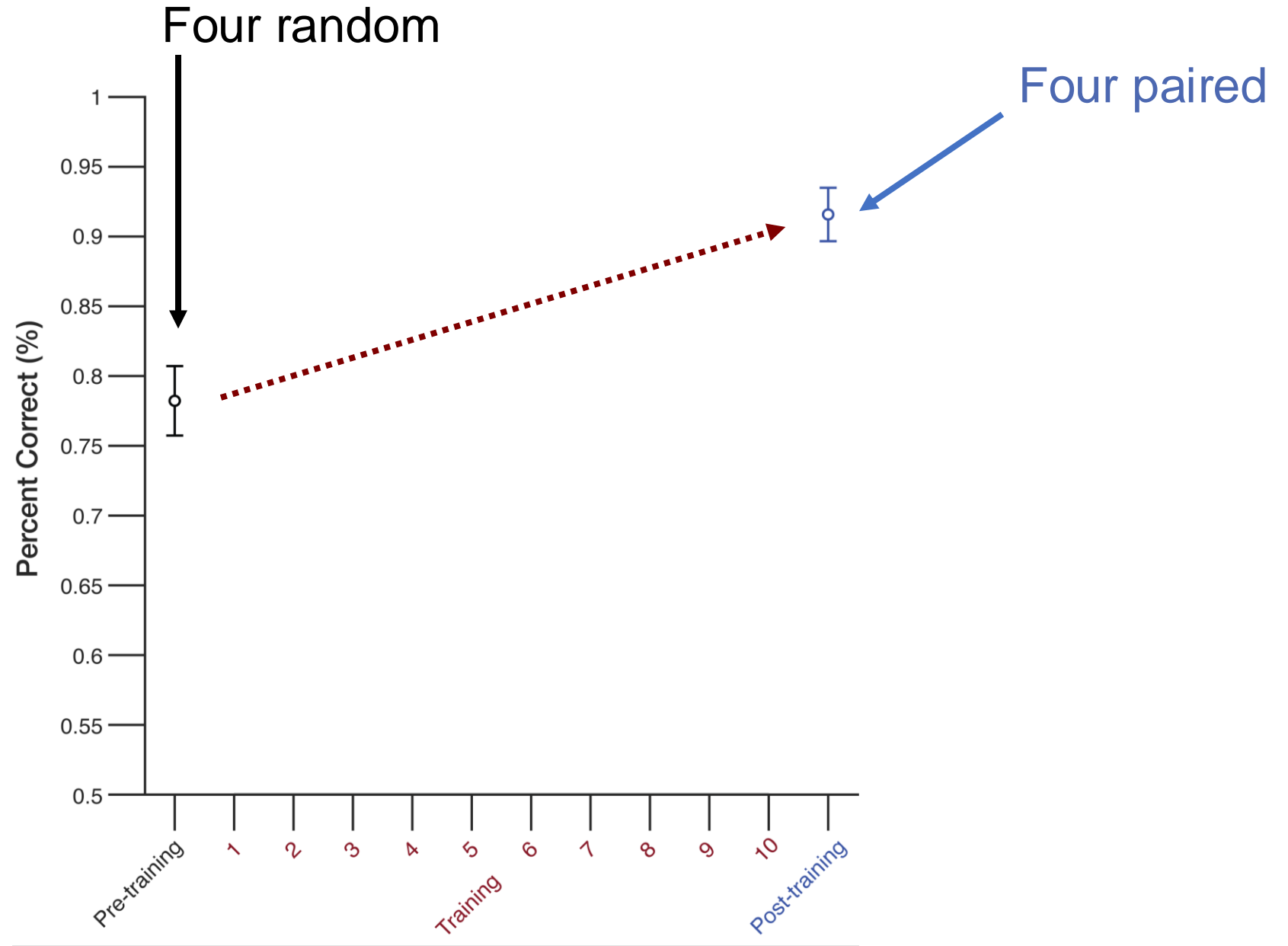
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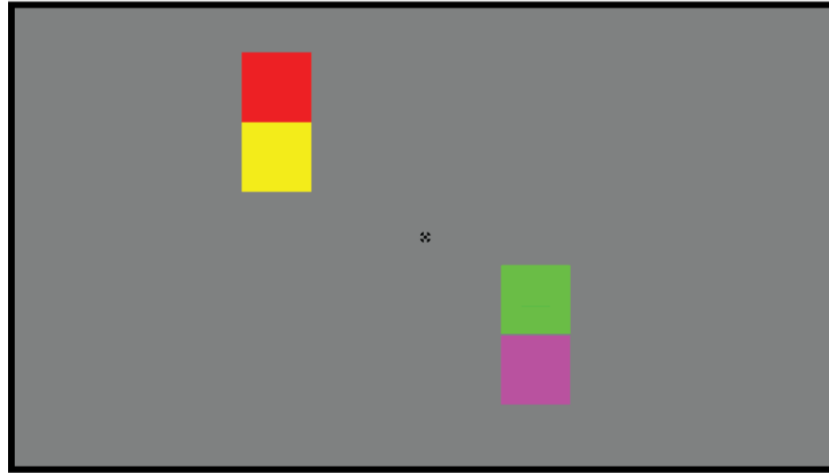
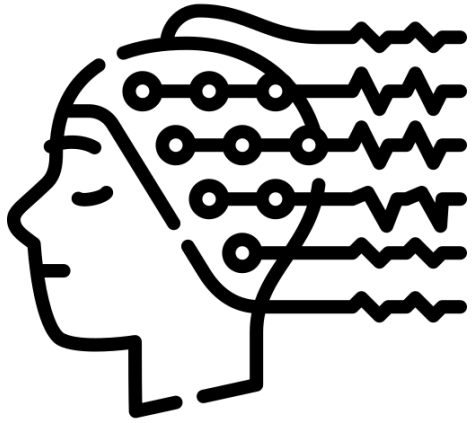
# E1: training session – aggregate performance



# E1: training session – average performance

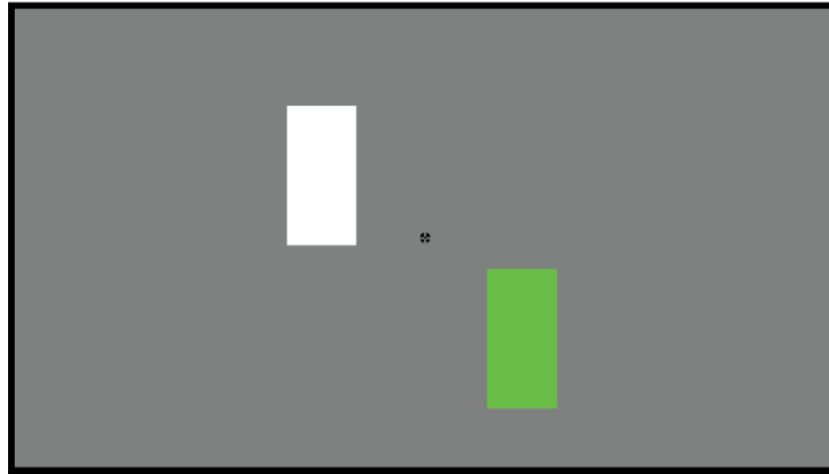


# Experiment 1: EEG session



Four random  
Four paired

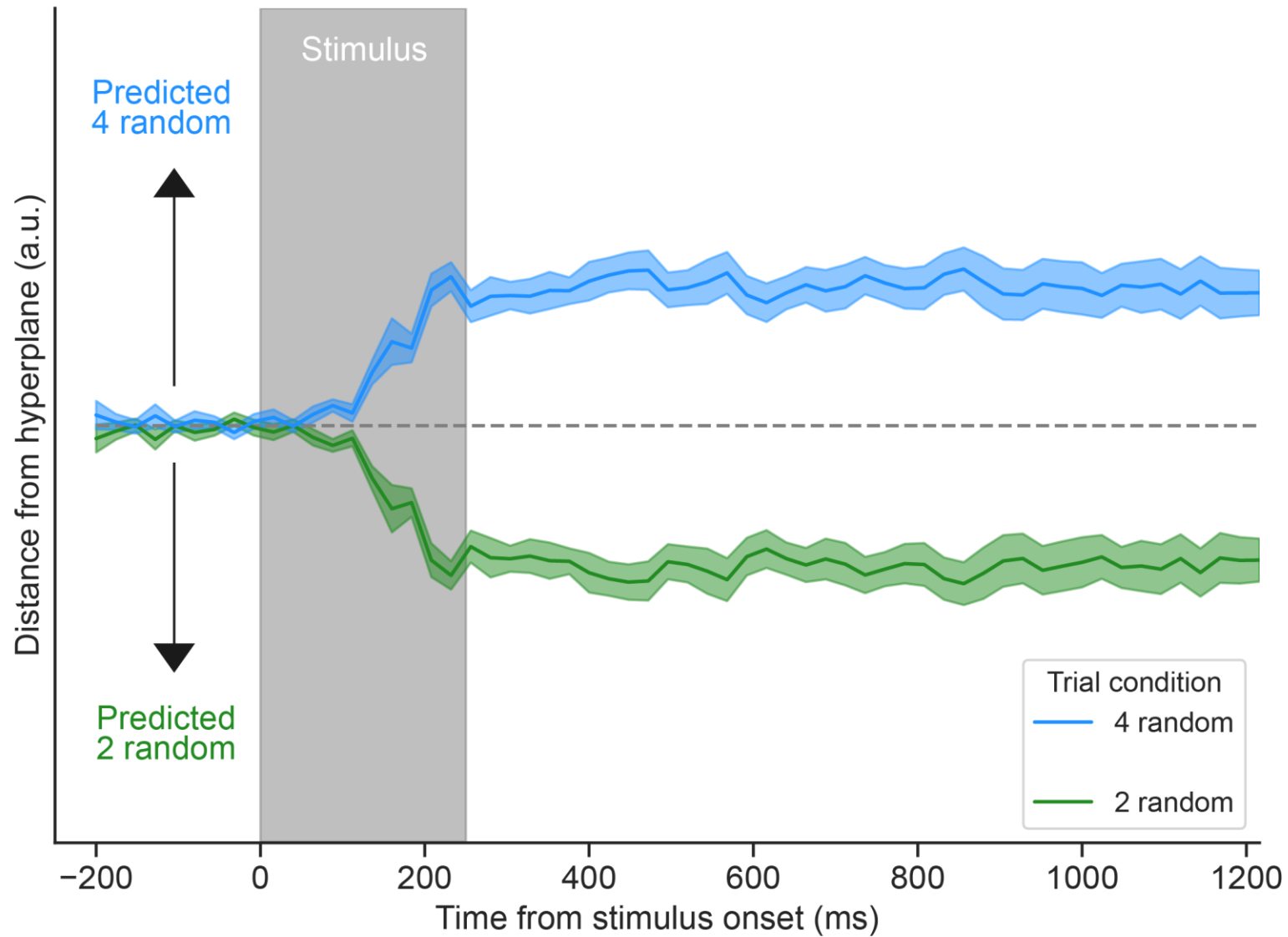
Perceptually  
equivalent



Two random

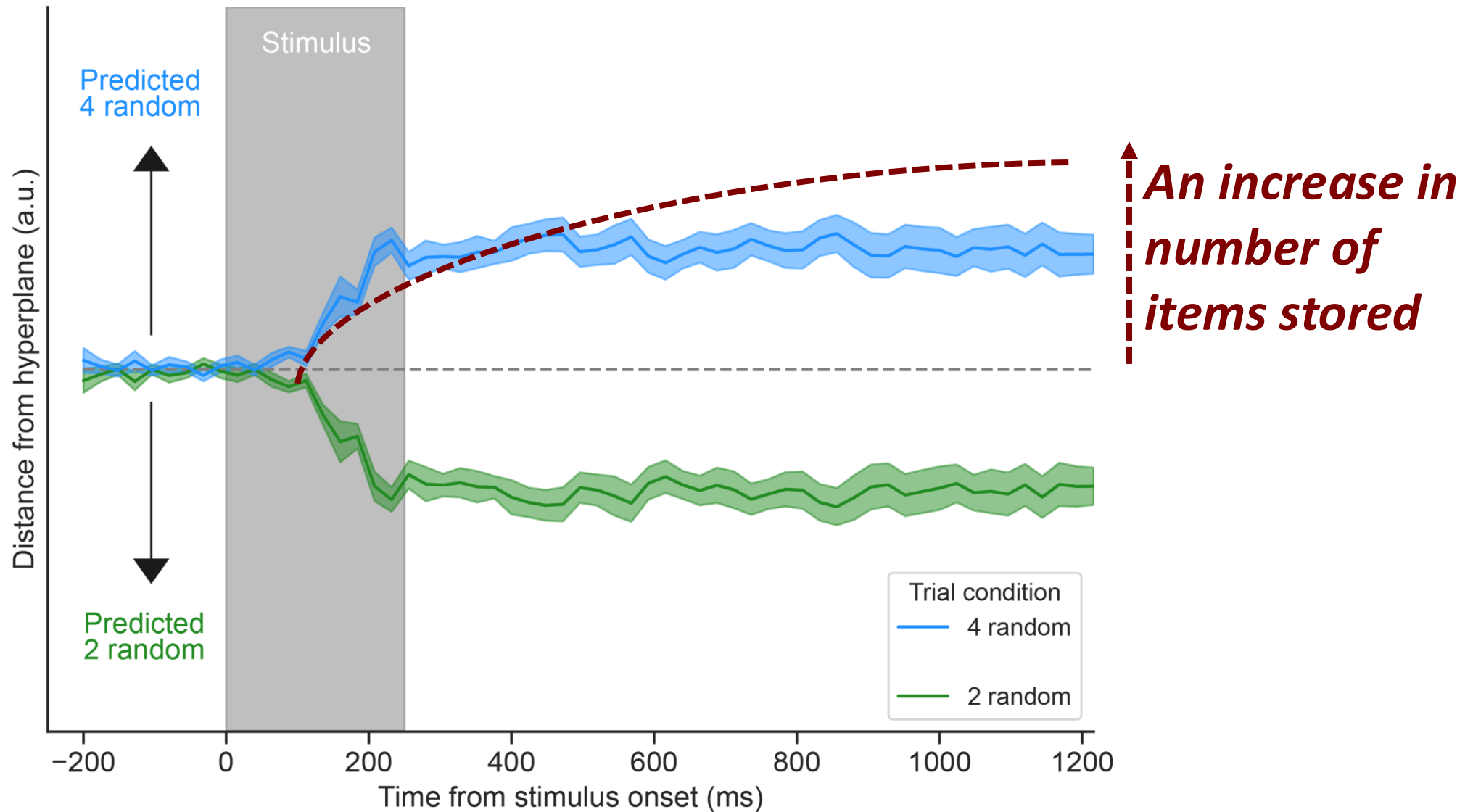


# Train 2 random versus 4 random

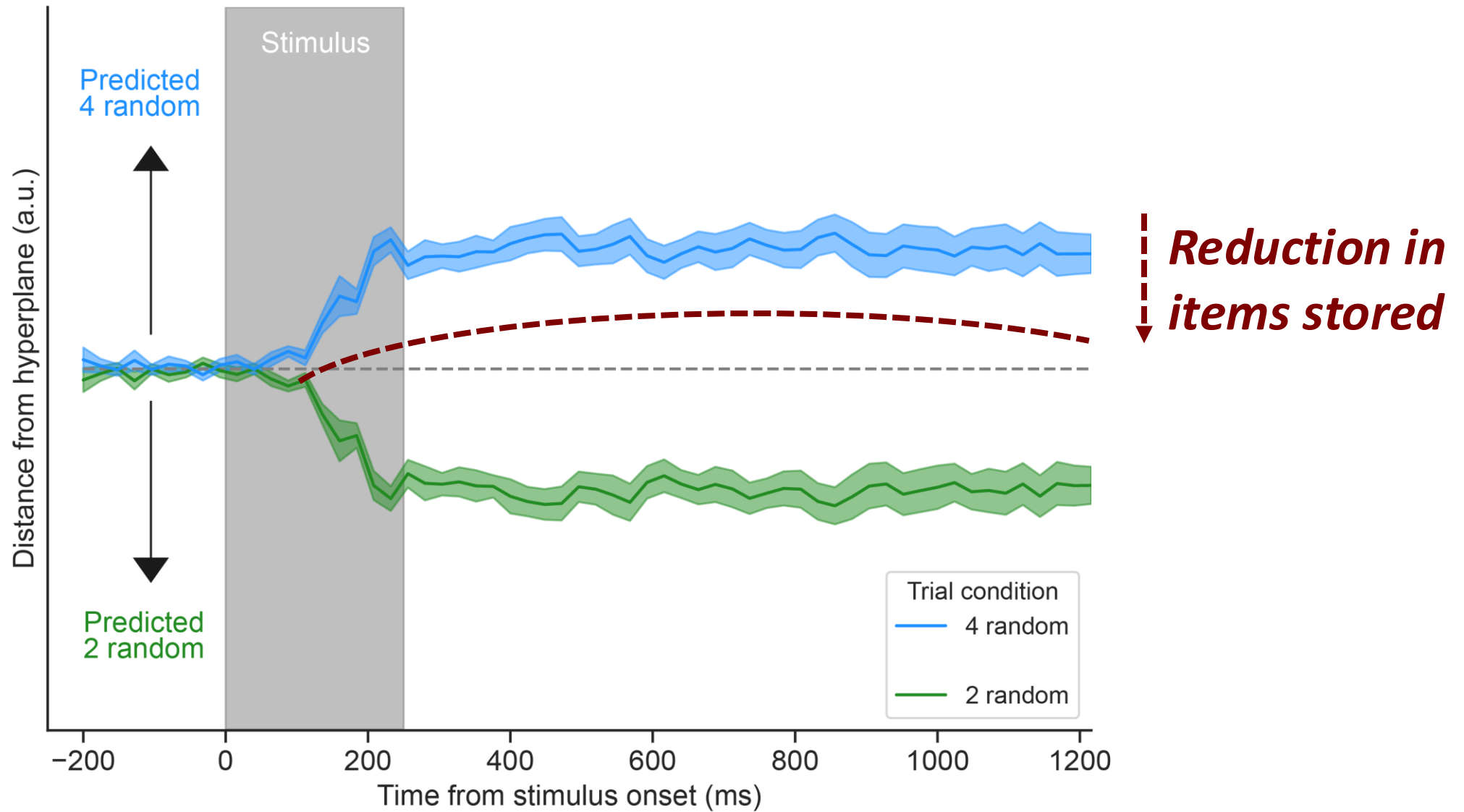




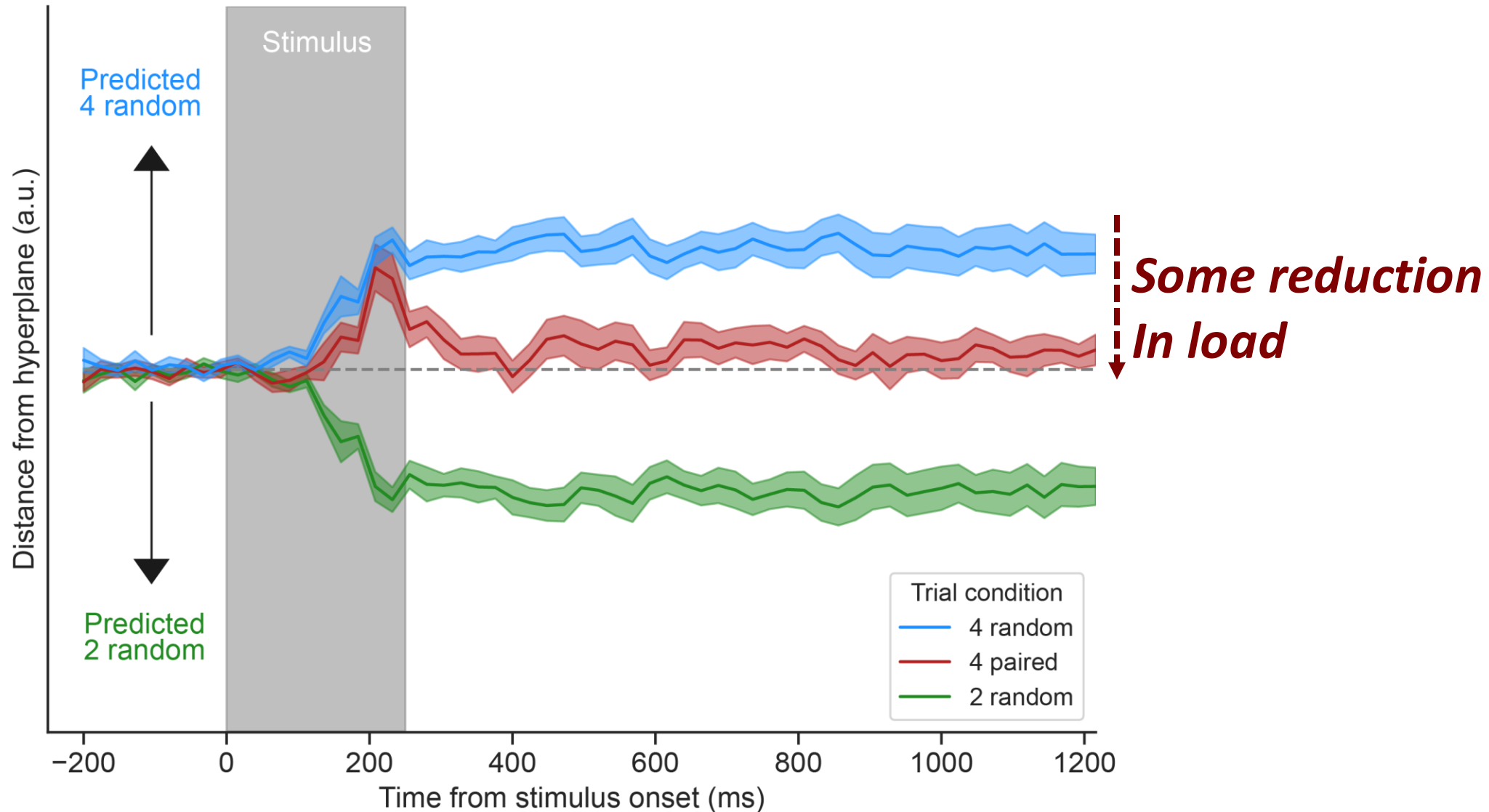
# Test on 4 paired?



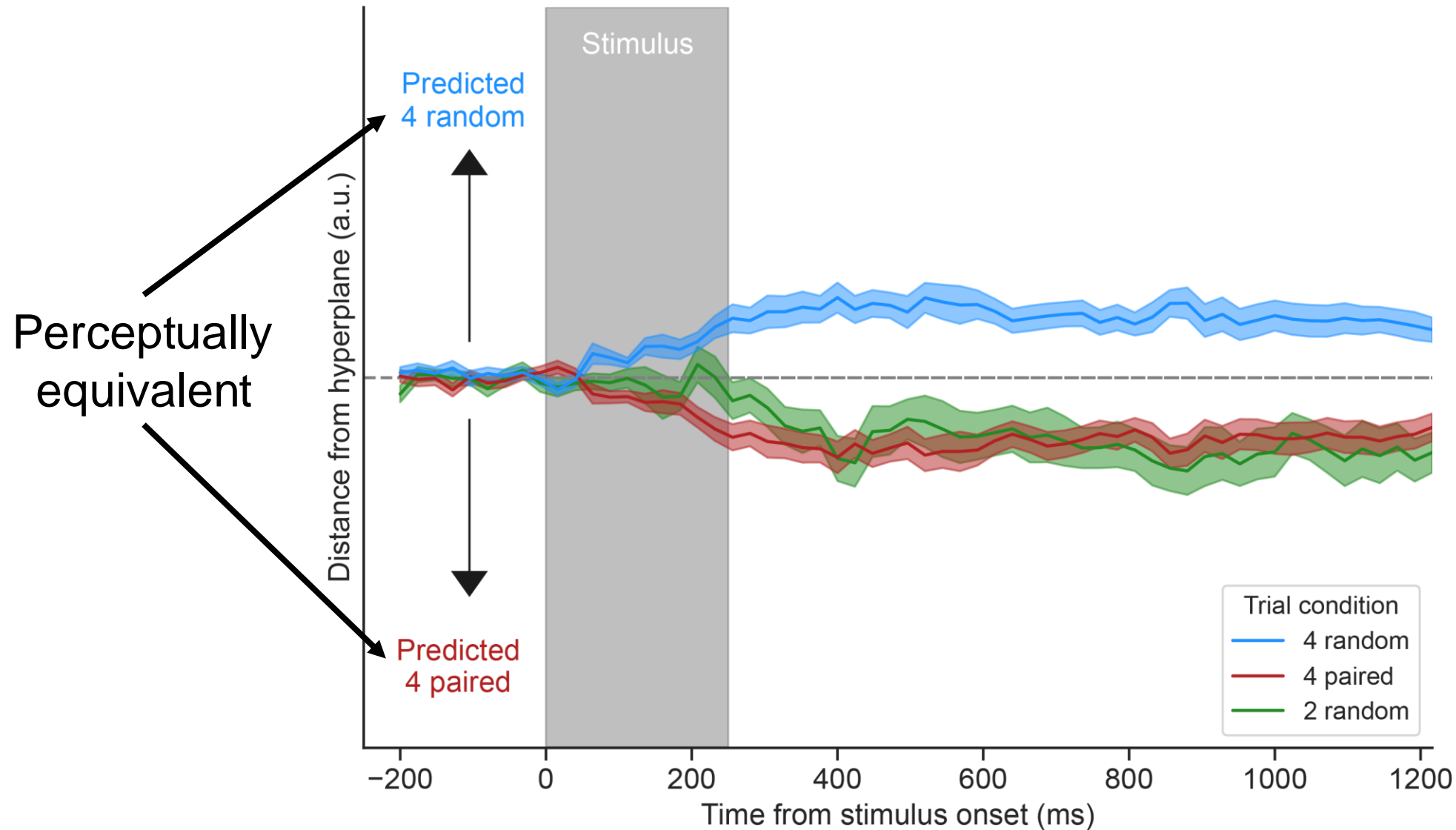
# Test on 4 paired?



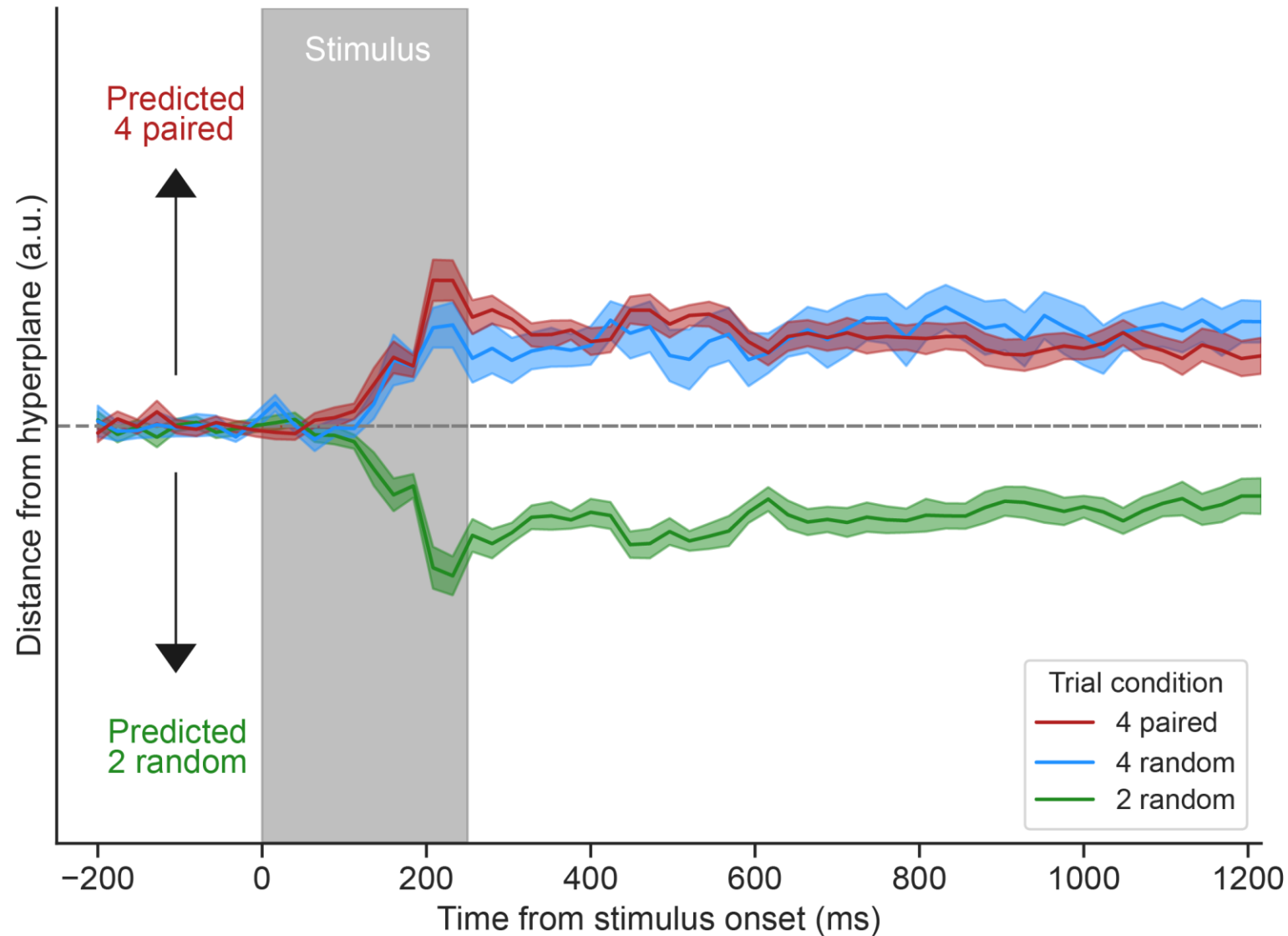
# Train 2 random versus 4 random, test 4 paired



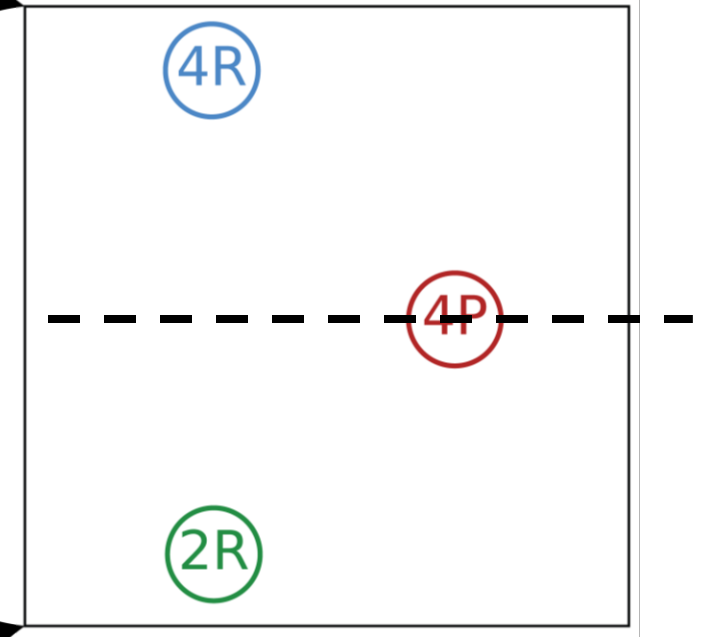
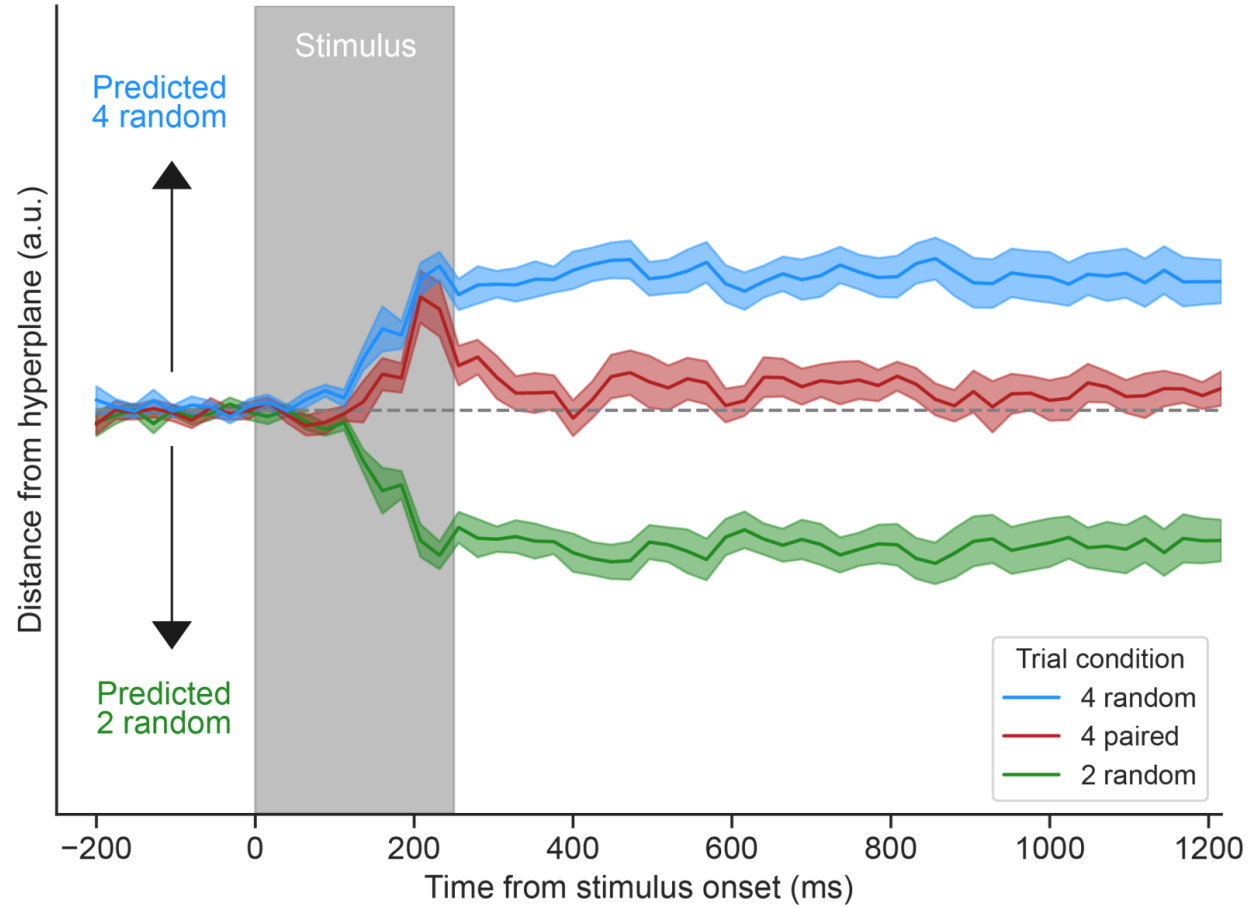
# Train 4 random versus 4 paired, test 2 random



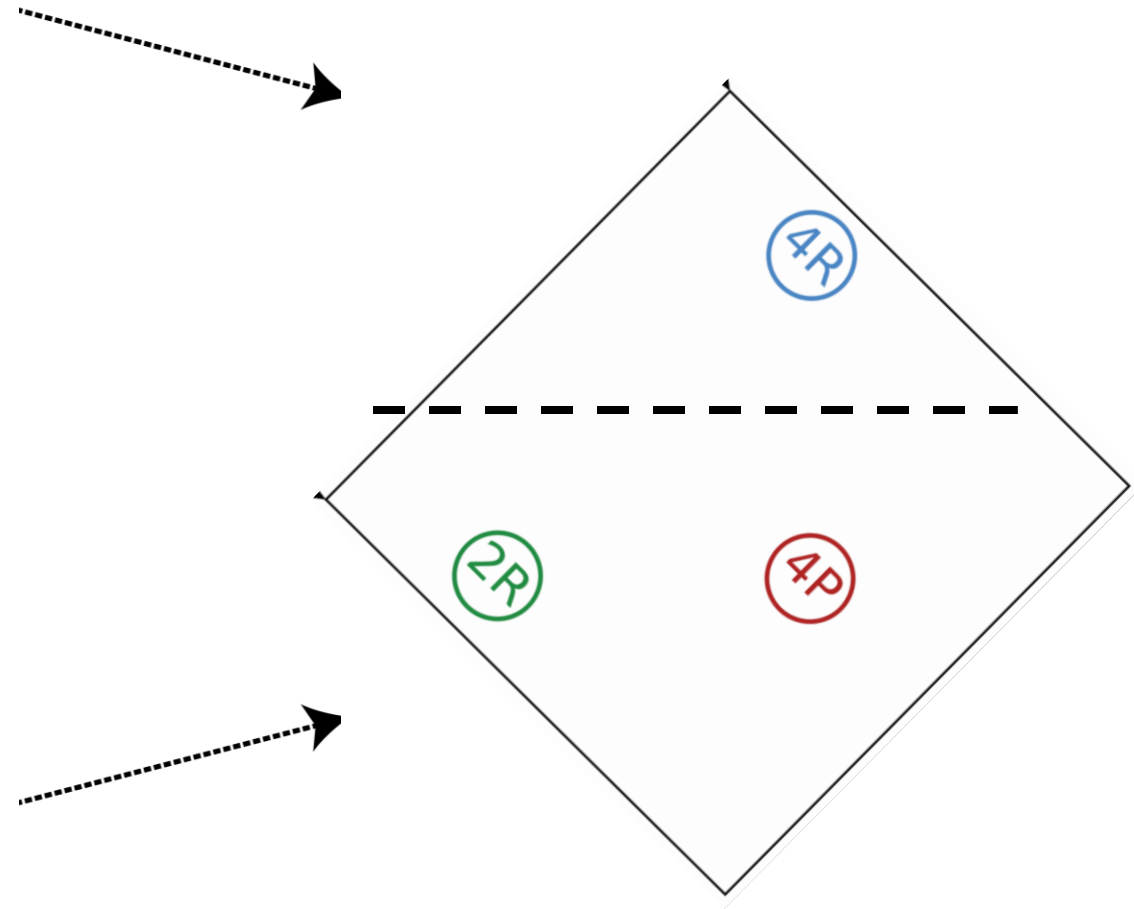
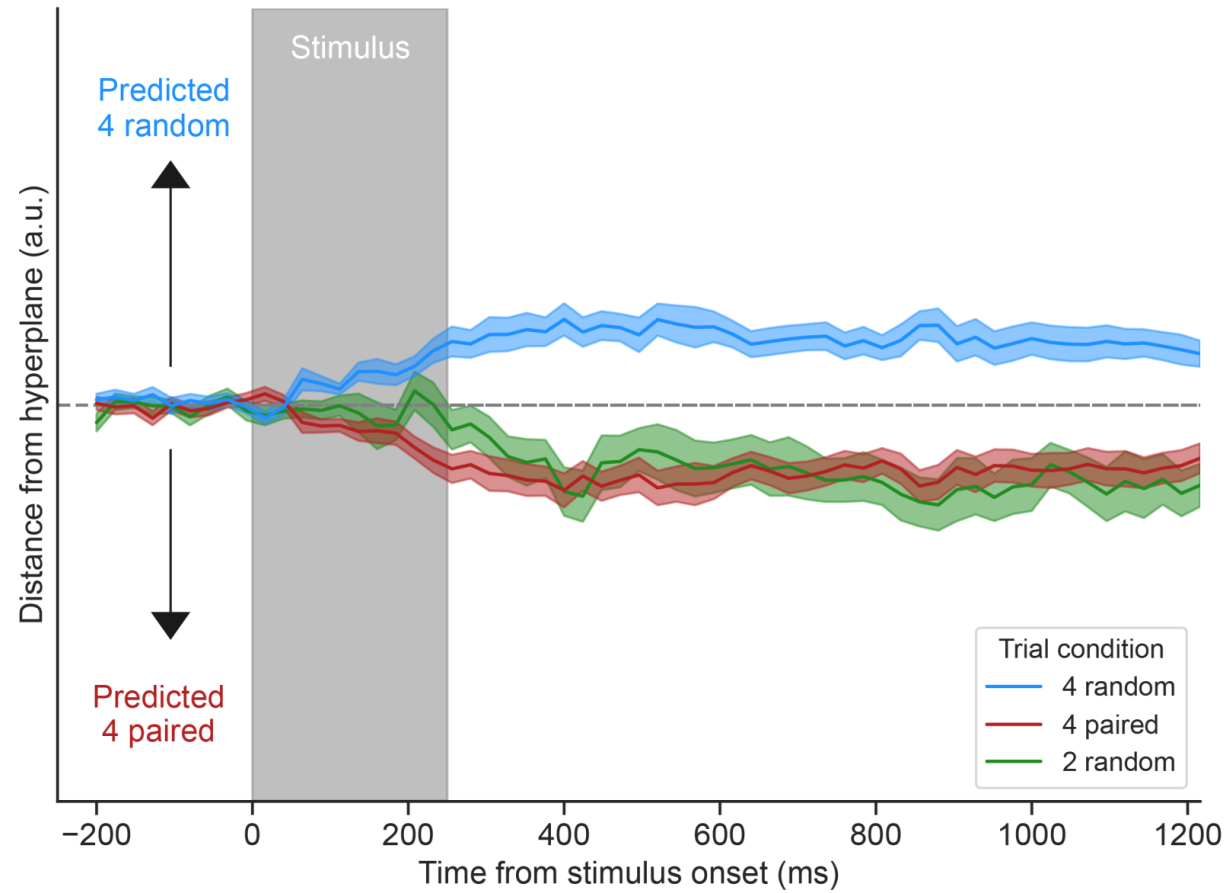
# Train 2 random versus 4 paired, test 4 random



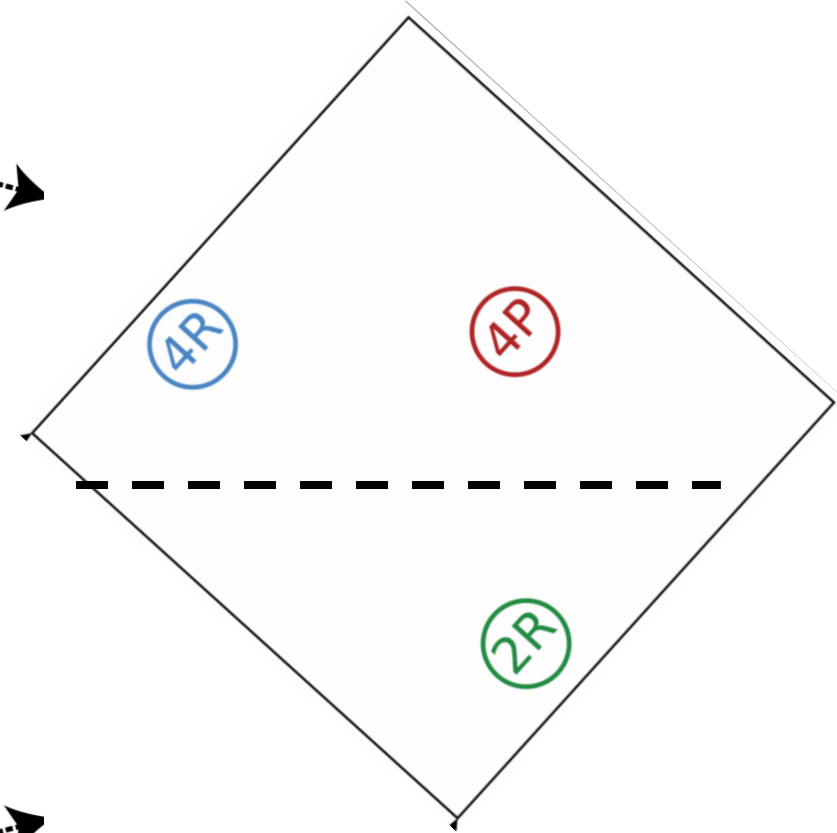
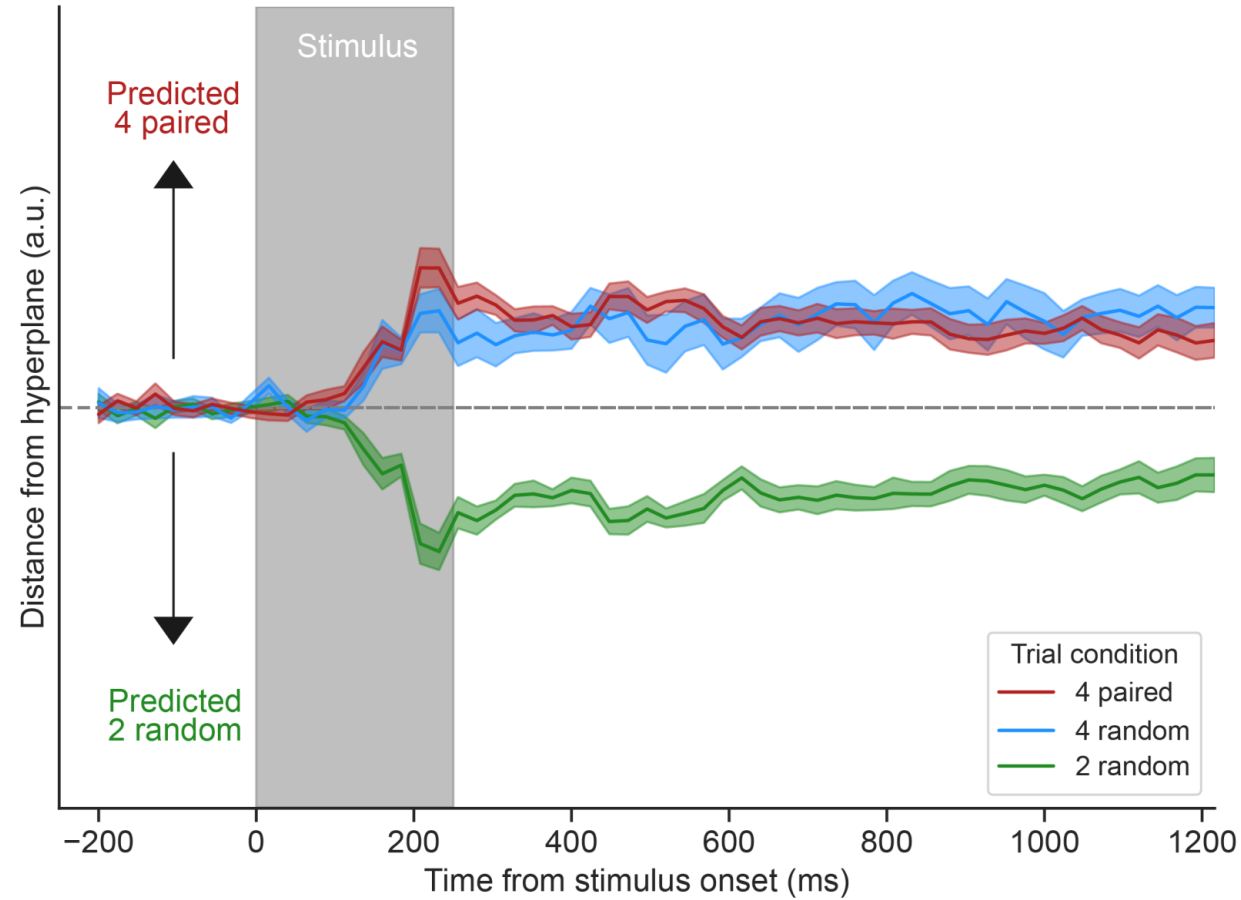
# Multidimensional scaling



# Multidimensional scaling



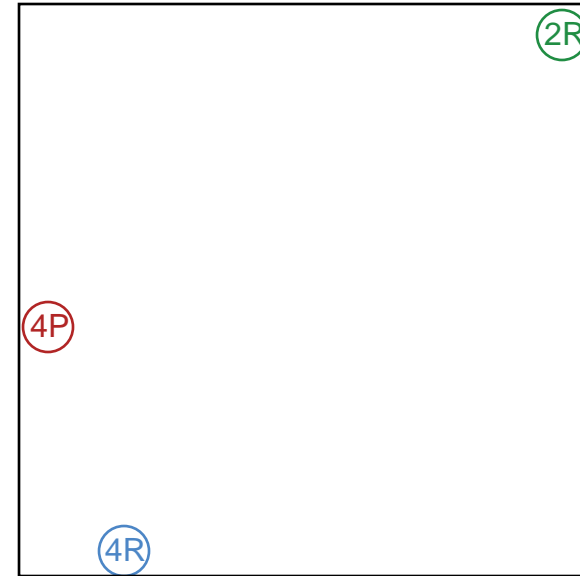
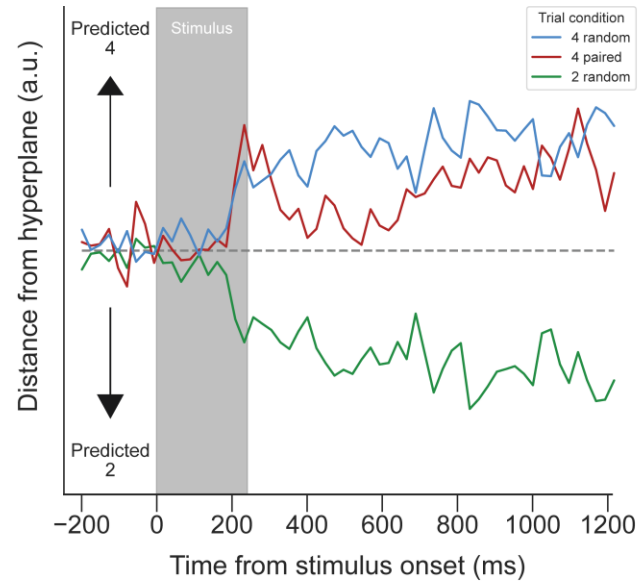
# Multidimensional scaling



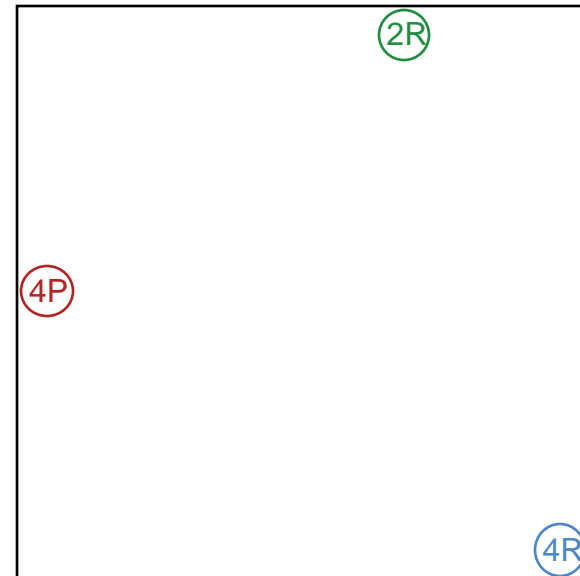
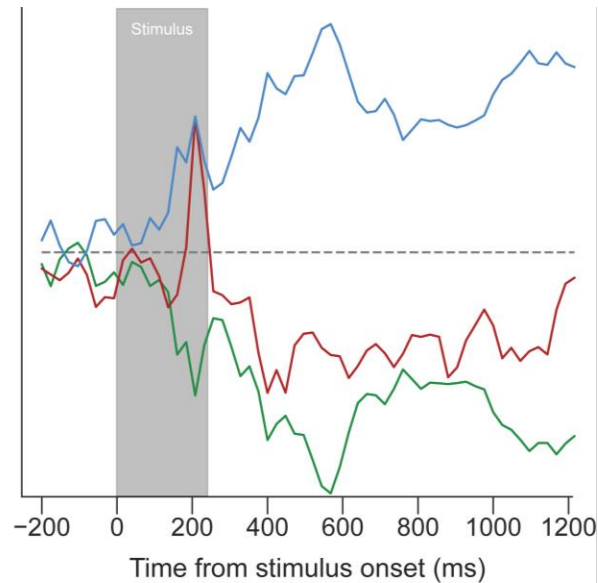


# Individual differences

“Weak chunking”

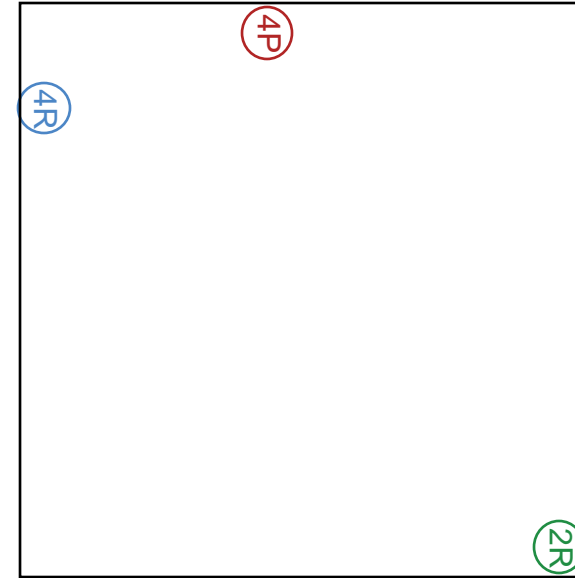
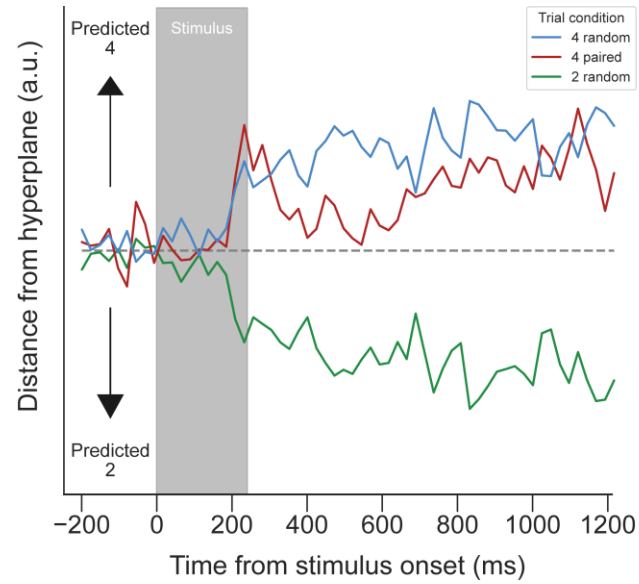


“Strong chunking”

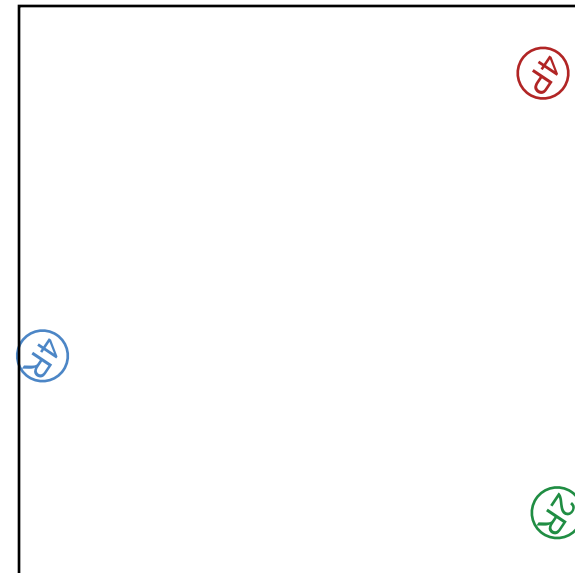
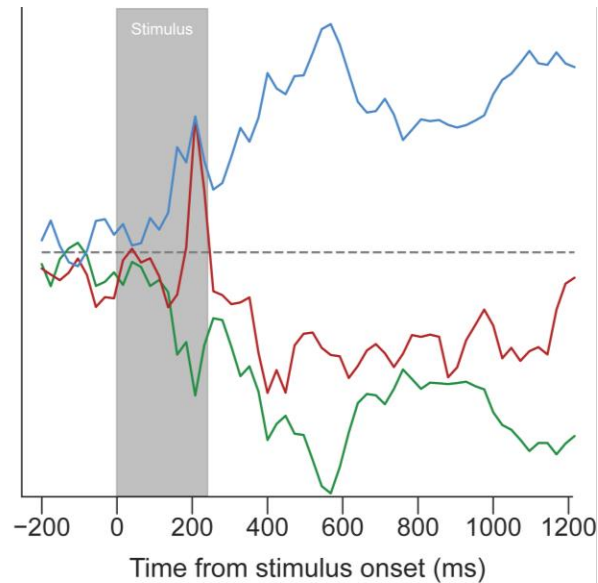


# Individual differences

“Weak chunking”

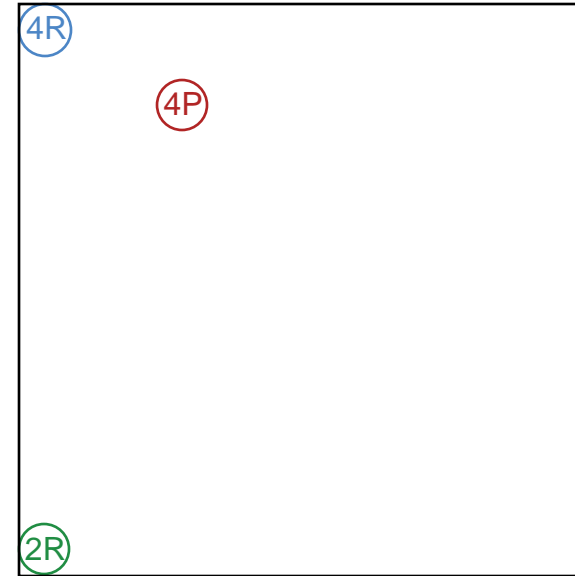
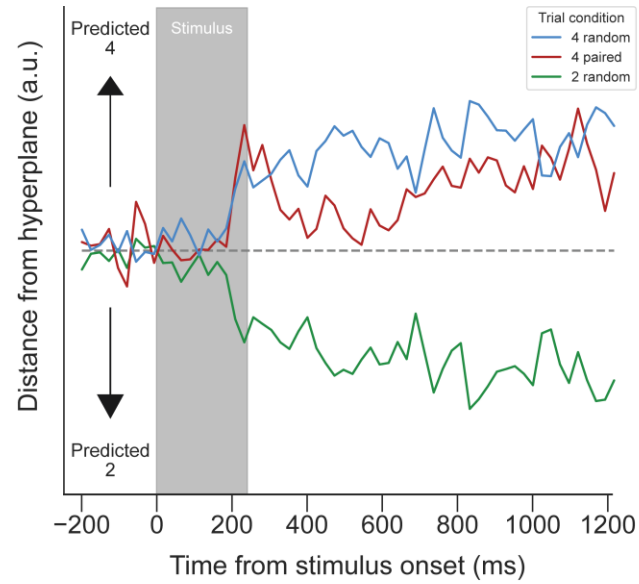


“Strong chunking”

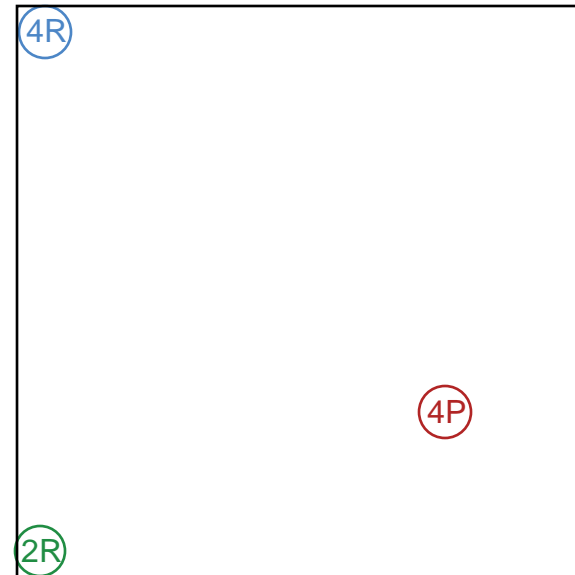
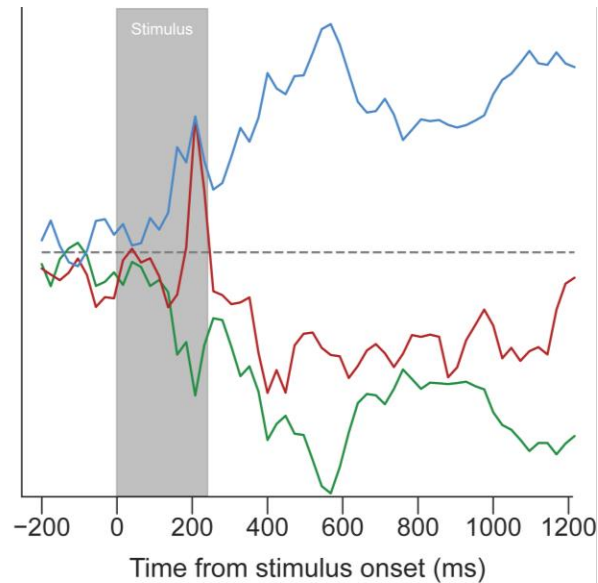


# Individual differences

“Weak chunking”

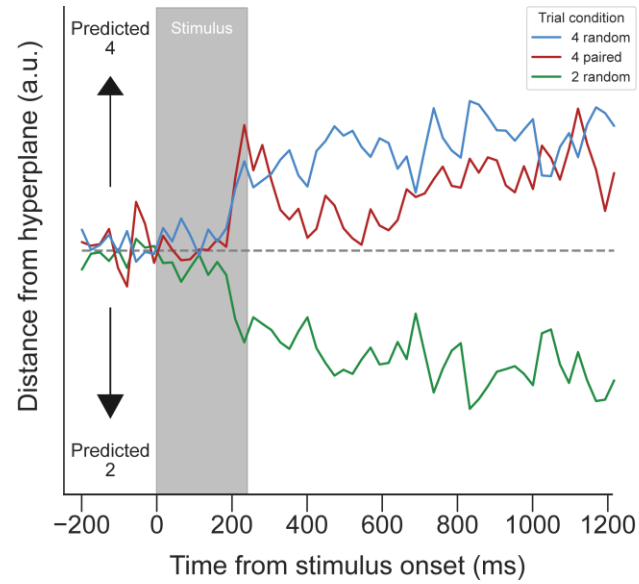


“Strong chunking”

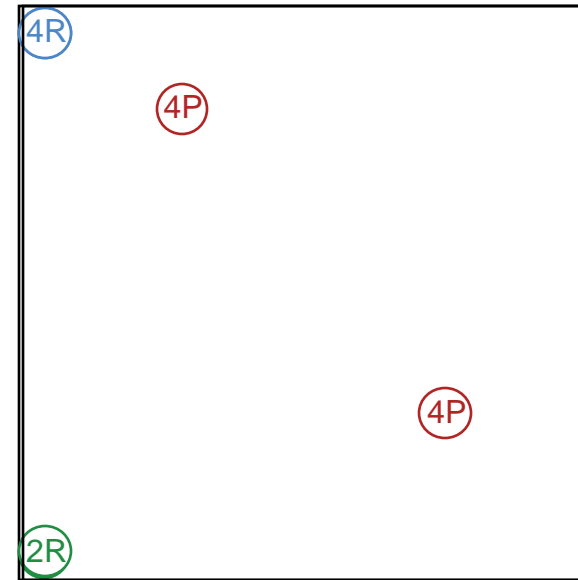
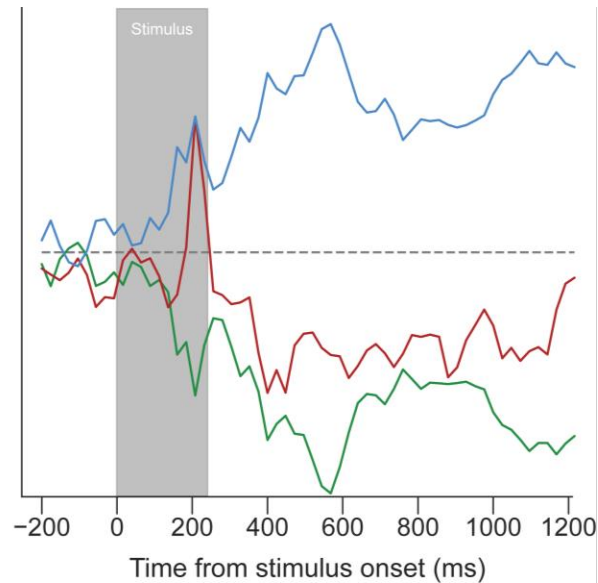


# Individual differences

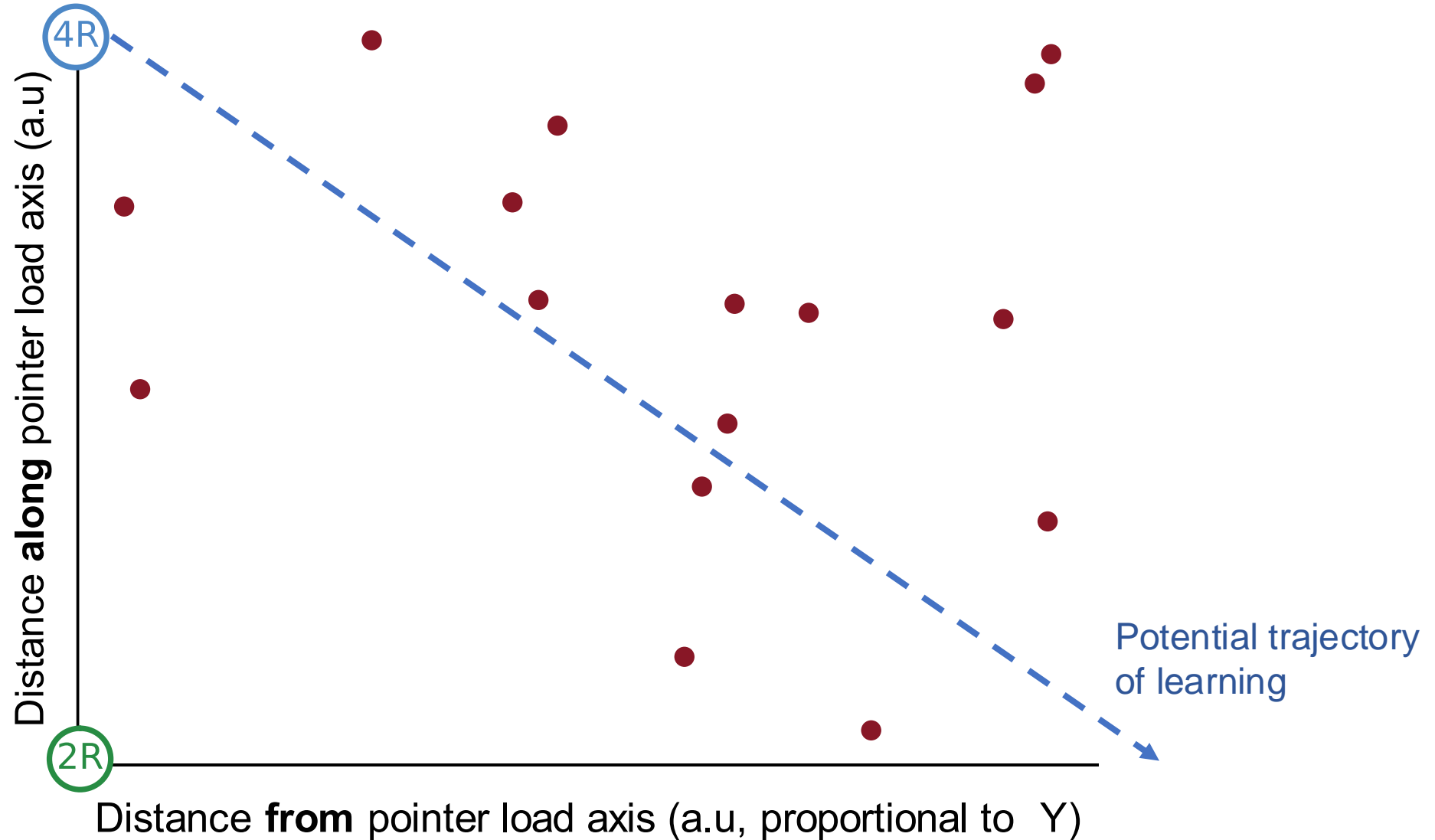
“Weak chunking”



“Strong chunking”



# Individual differences



# Training

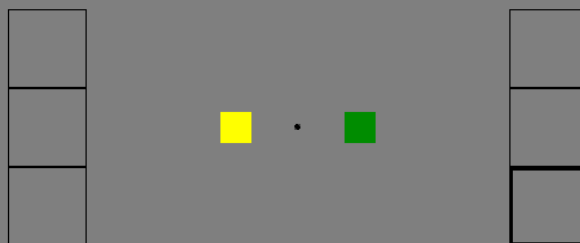
- Trained subjects to learn three color triplets



# Training

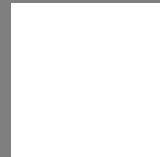
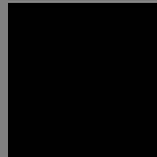
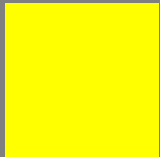
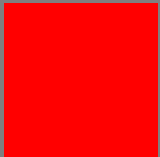
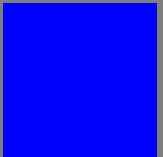


# Training



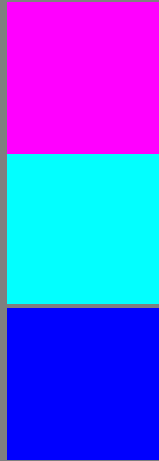


# Awareness Test

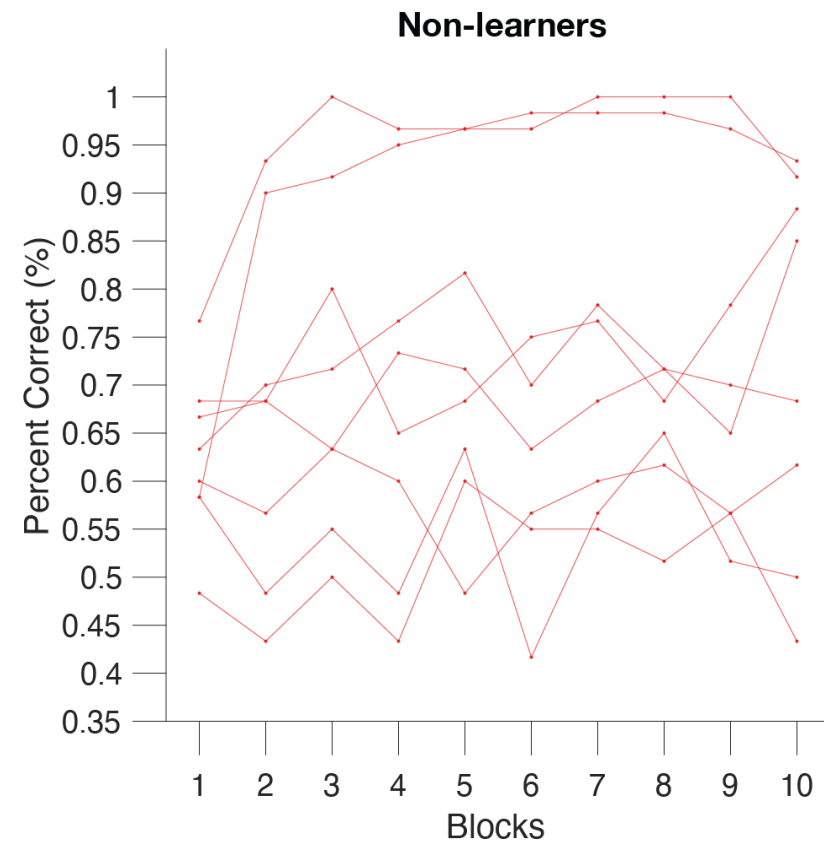
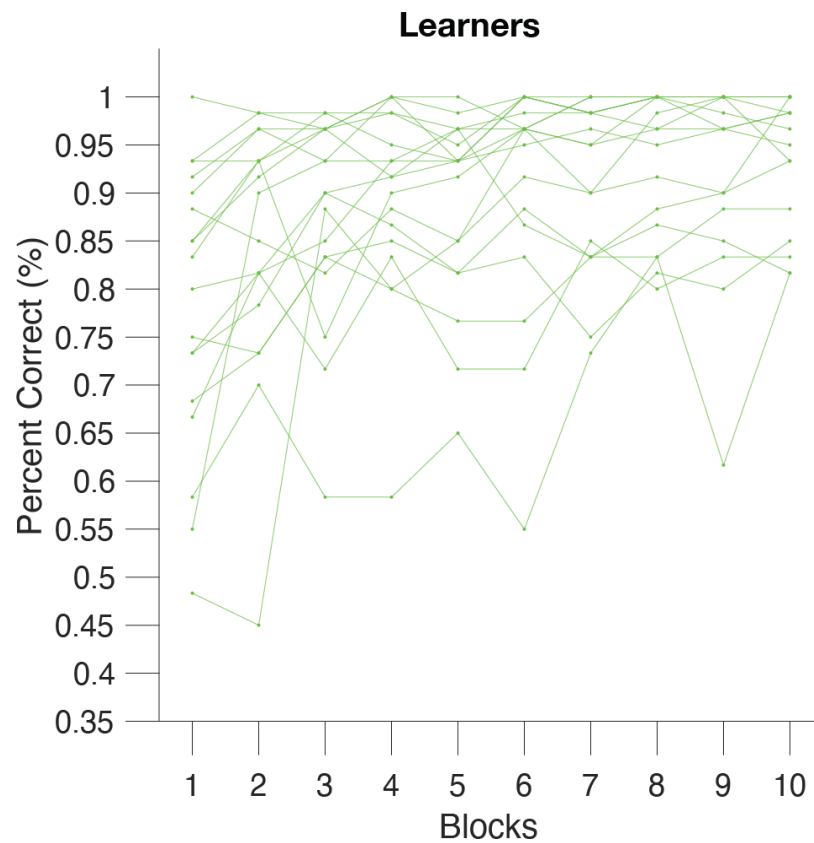
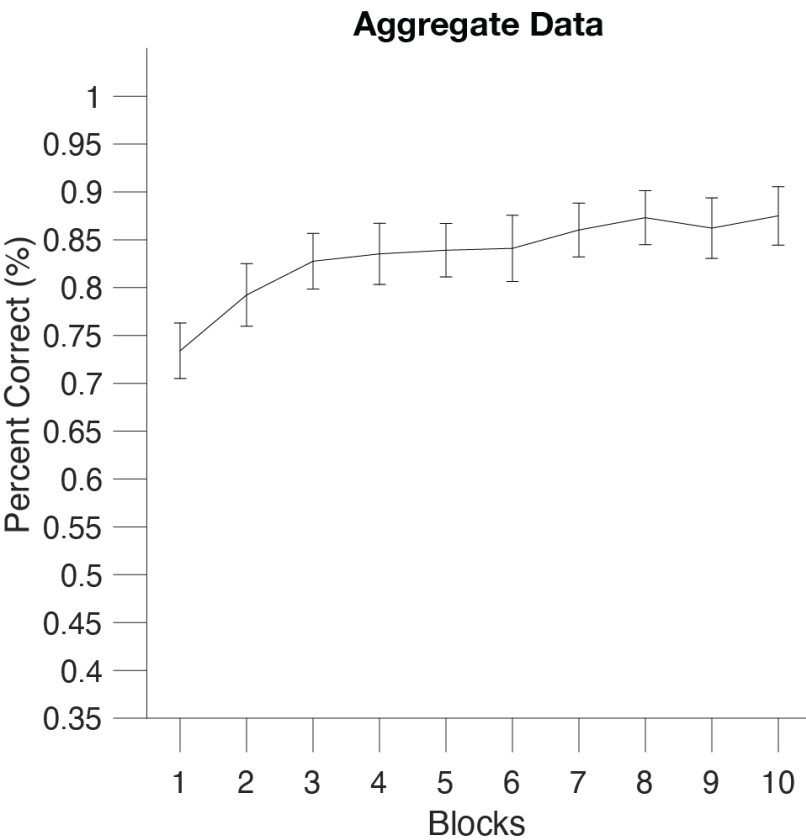


# Awareness Test

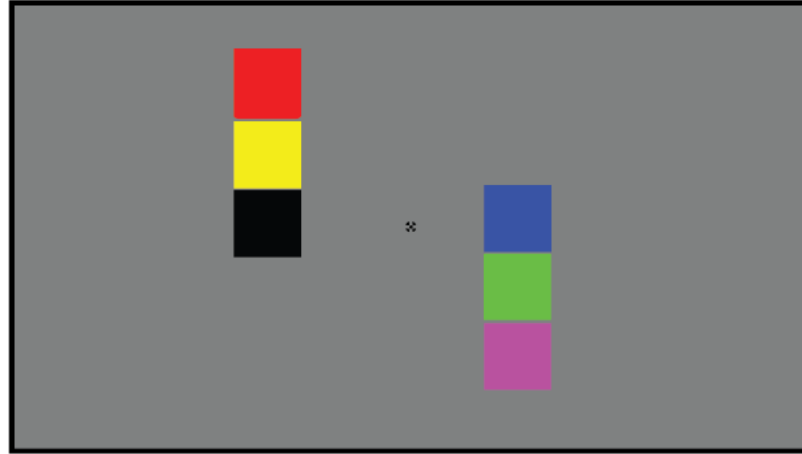
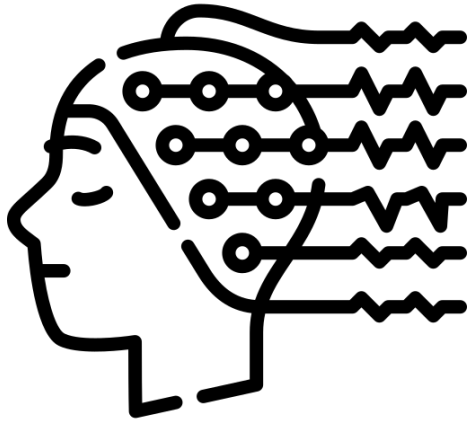
- Only subjects who correctly produced all triplets were considered “learners”



# Training Results

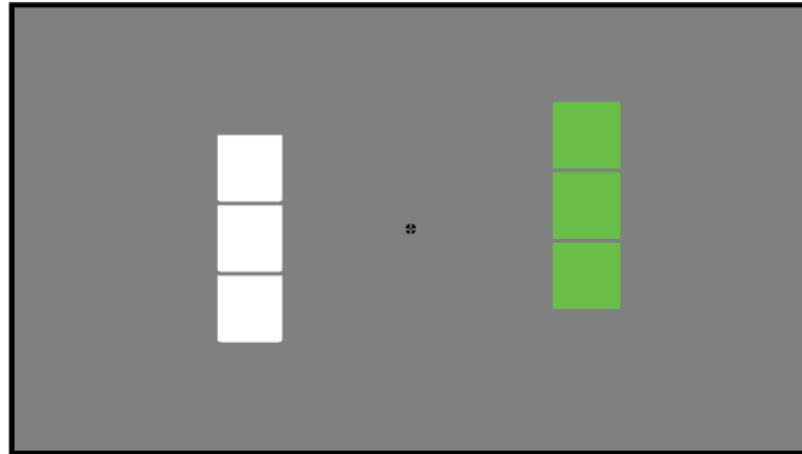


# EEG session



Six random  
Six chunked

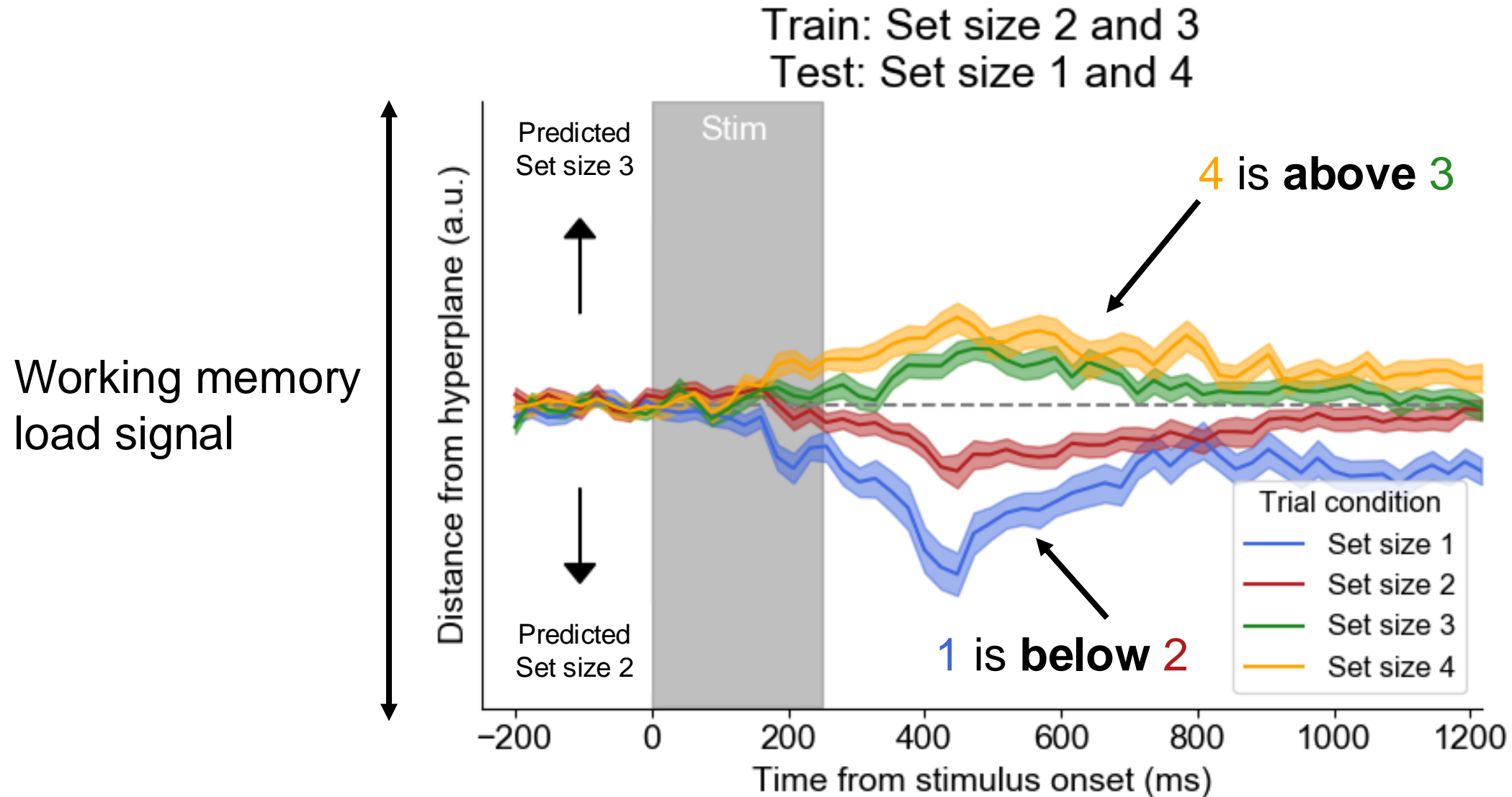
Perceptually  
equivalent



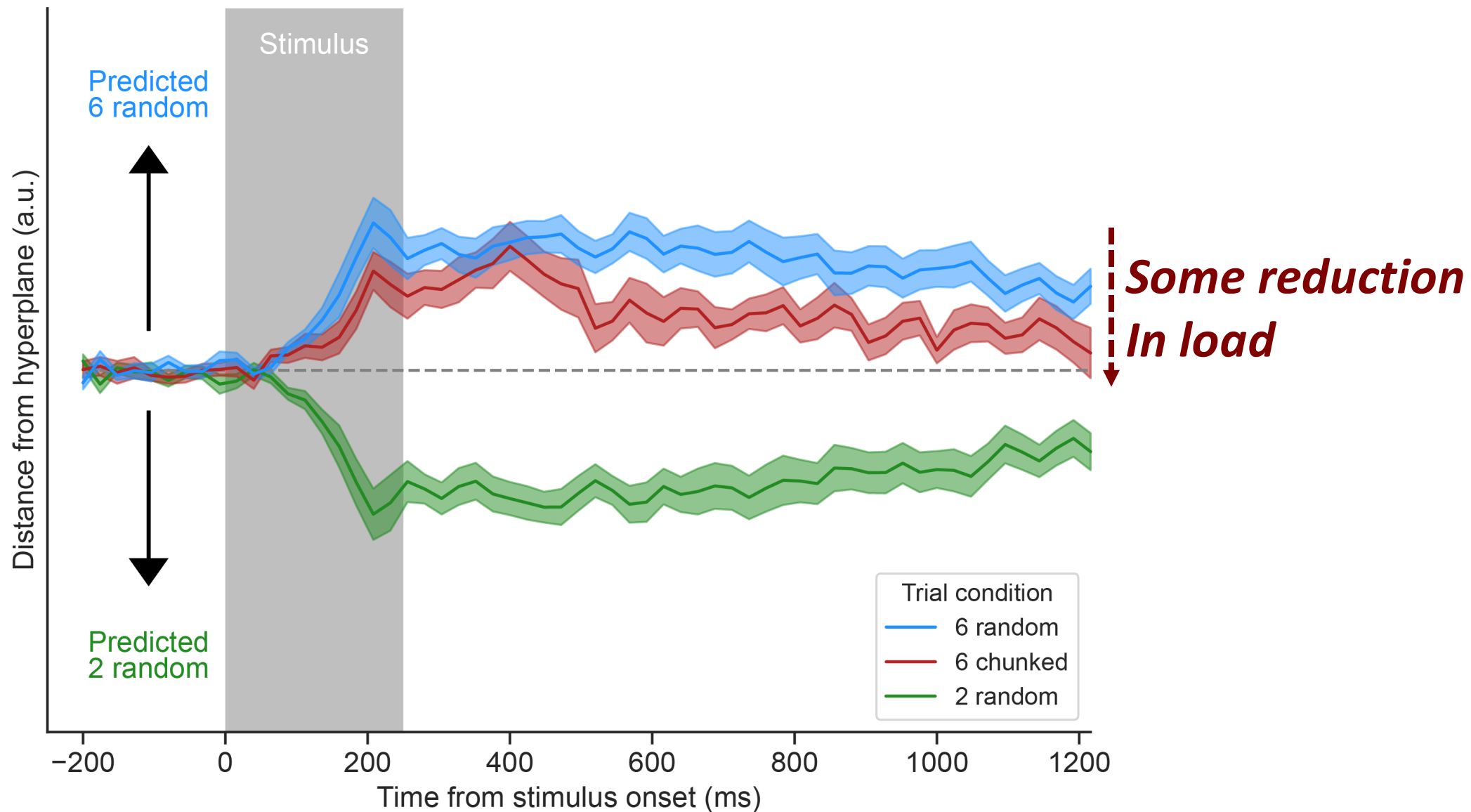
Two random



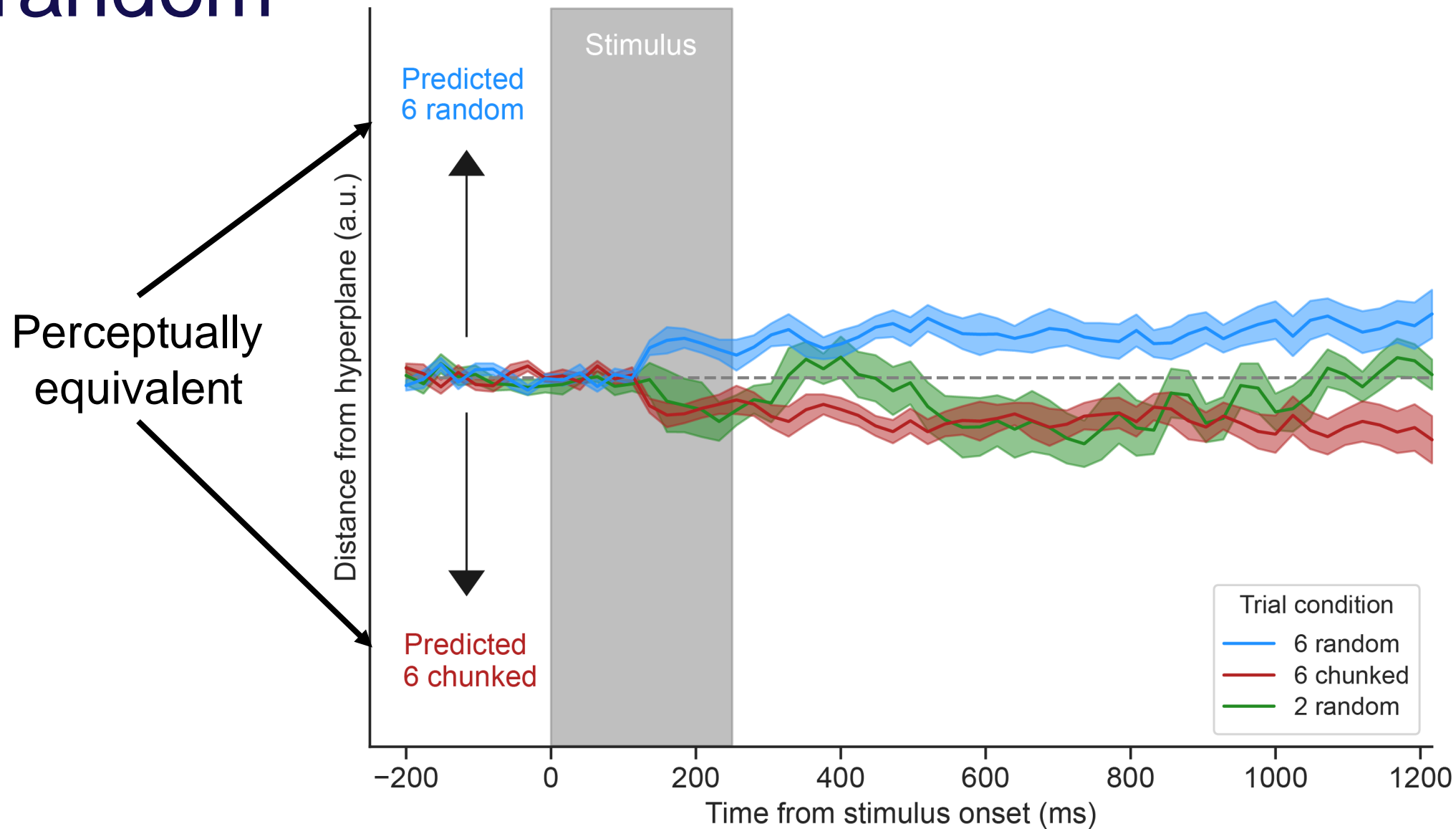
# Multivariate classification of working memory



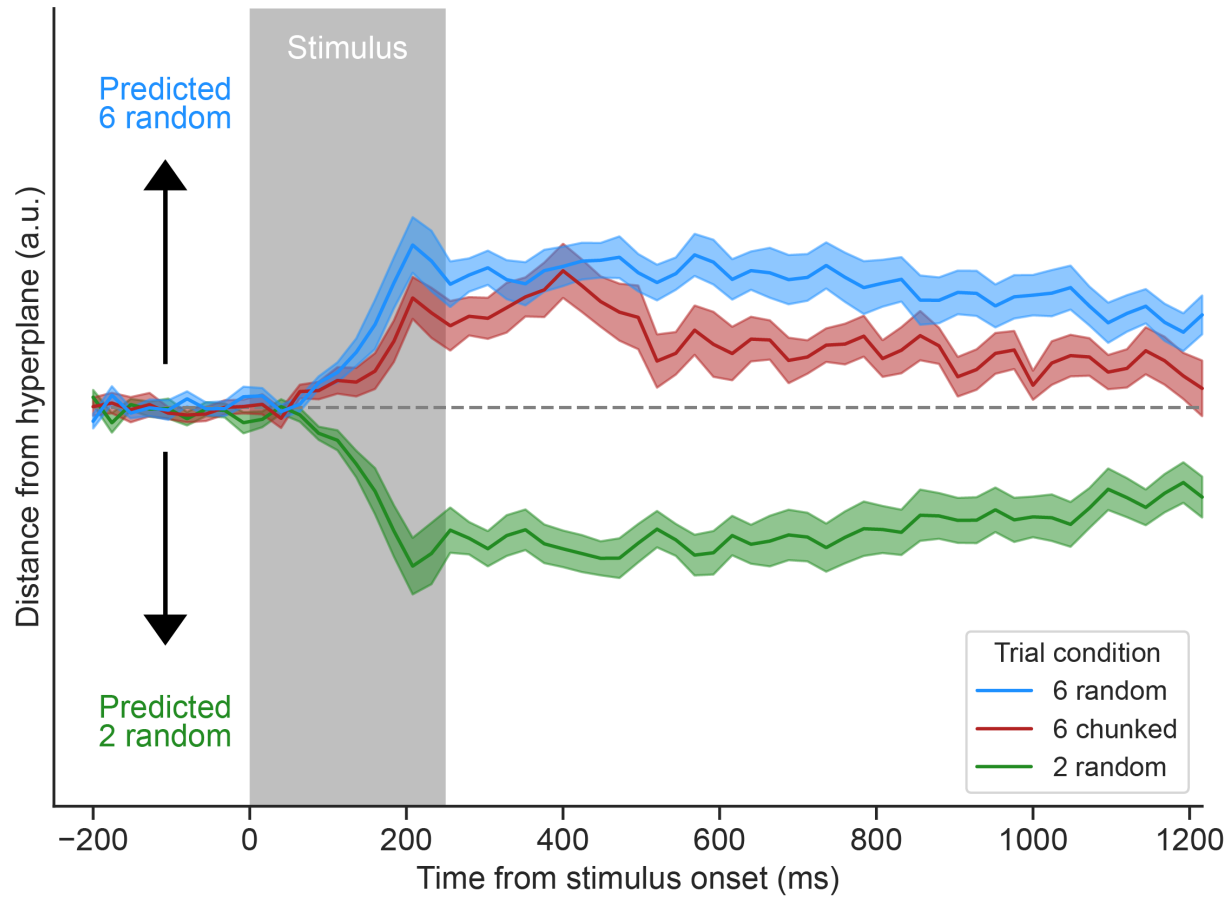
# Train 6 random versus 2 random, test 6 chunked



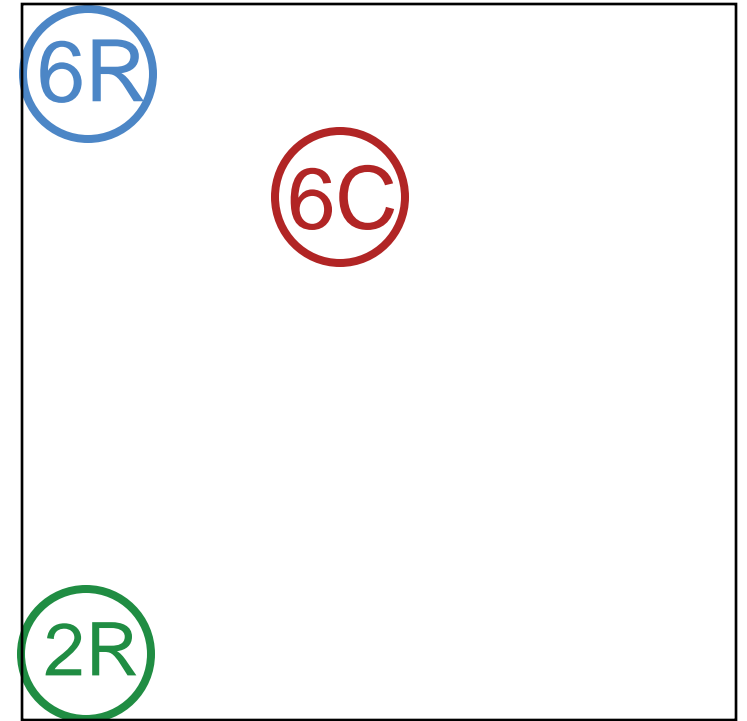
# Train 6 random versus 6 chunked, test 2 random



# Train 6 random versus 2 random, test 6 chunked

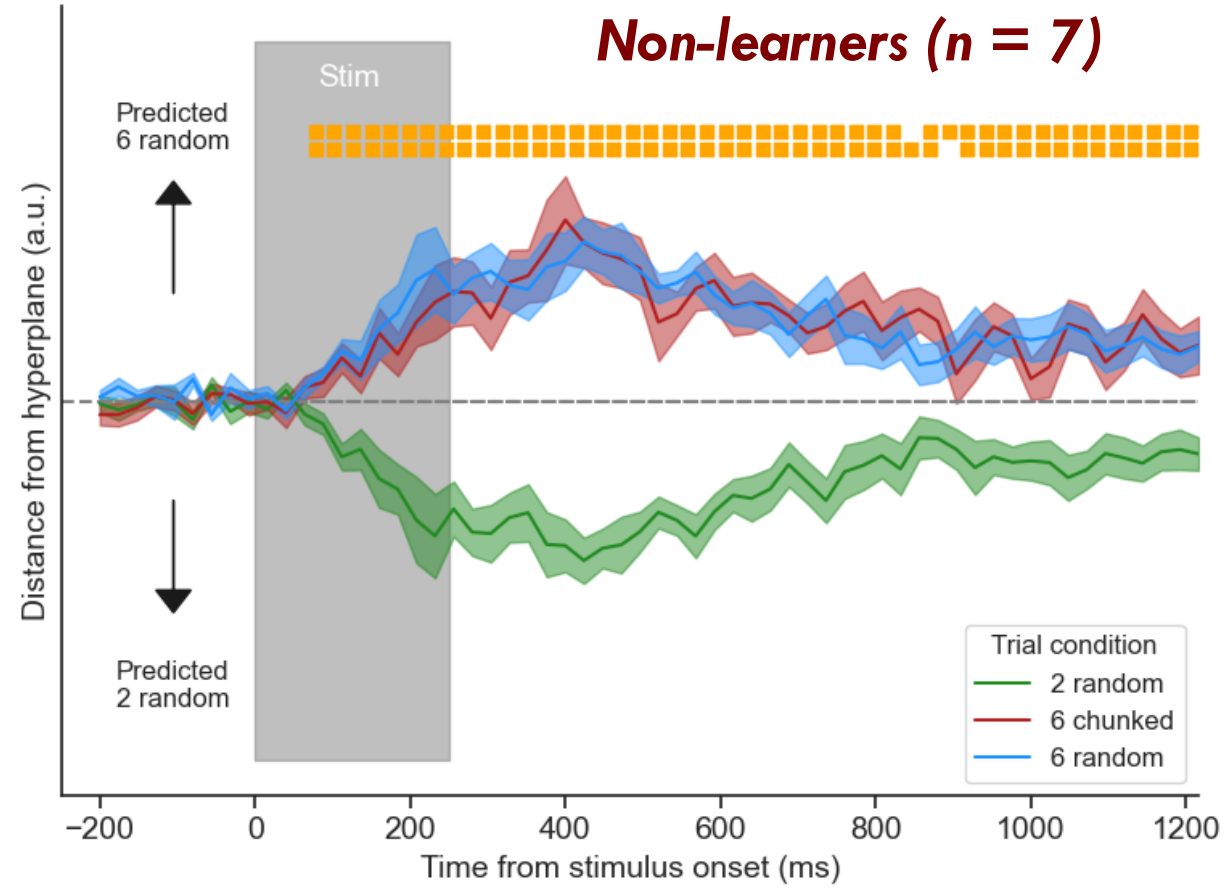
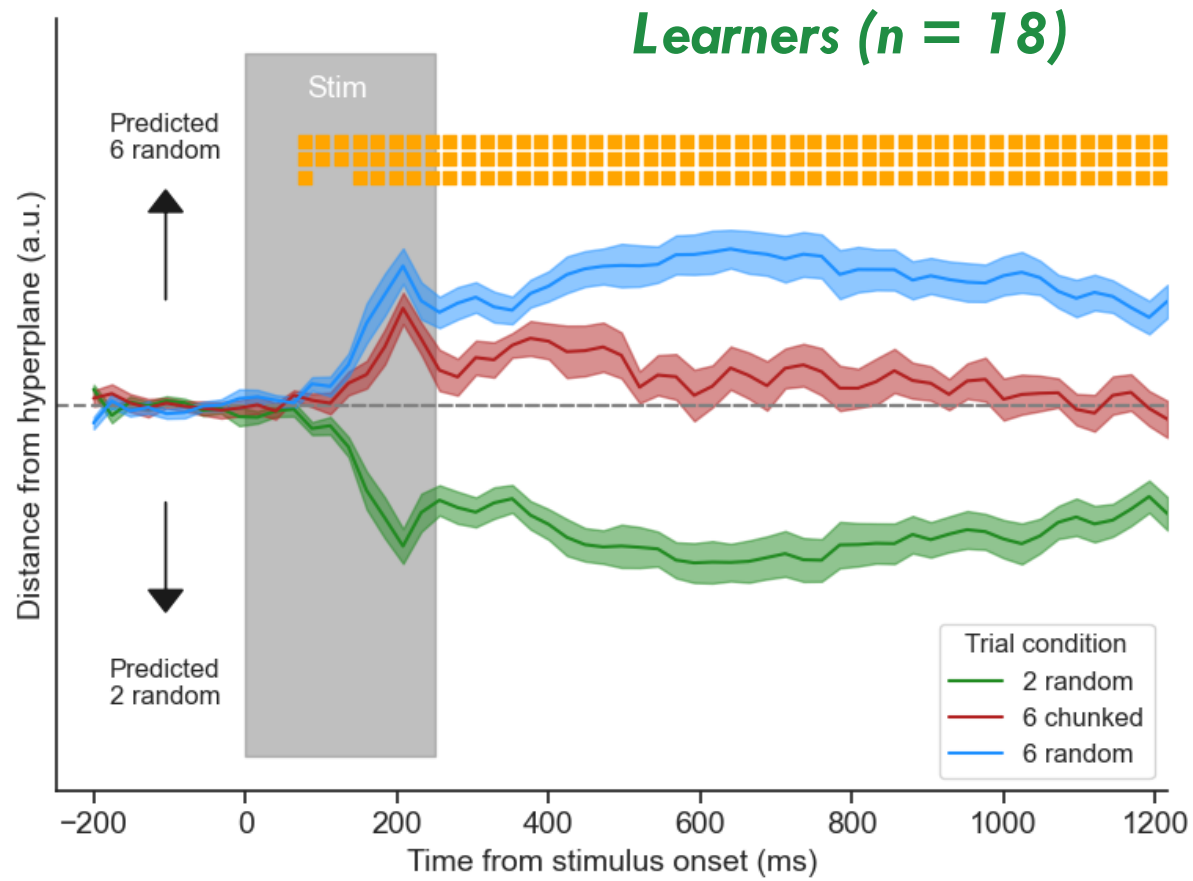


Multidimensional scaling



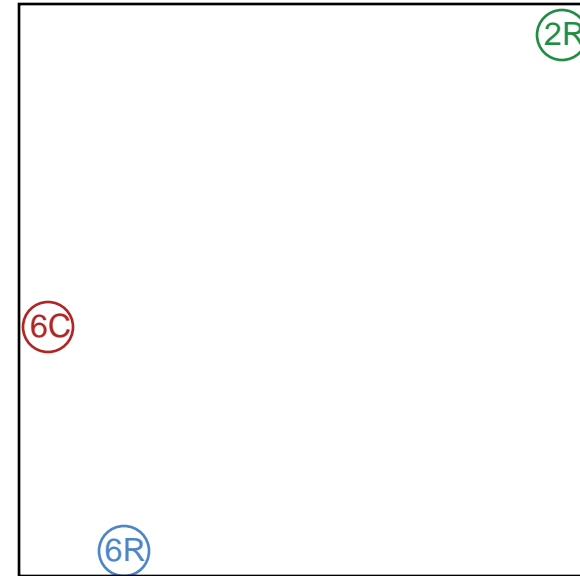
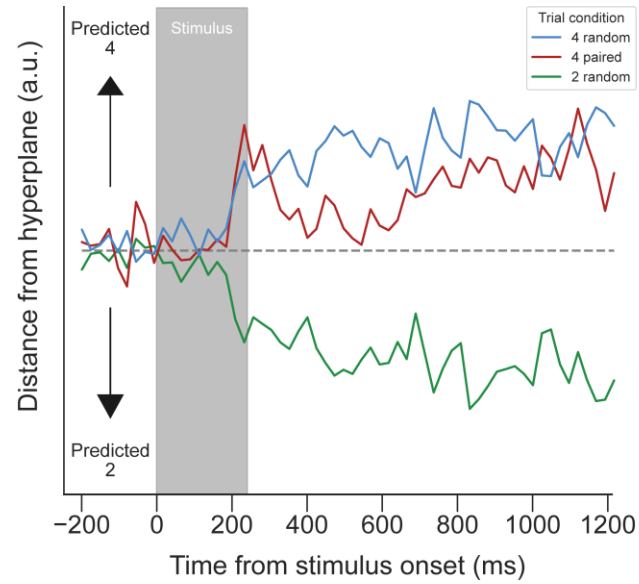


# Learners vs non-learners

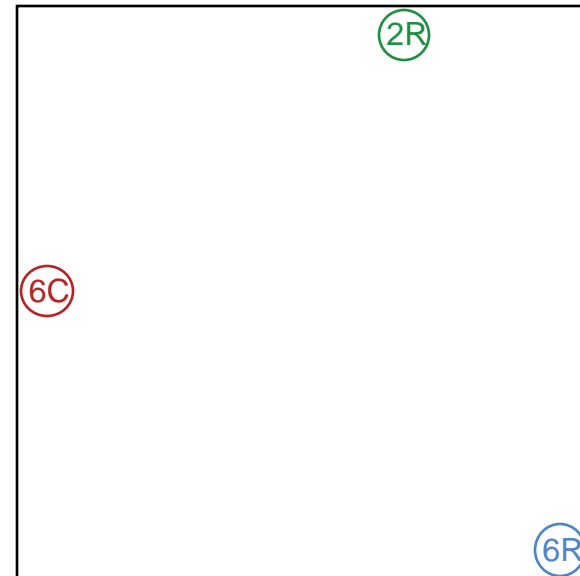
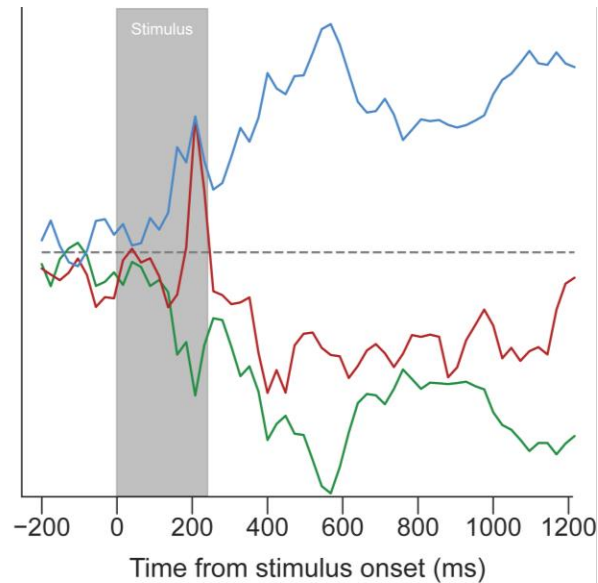


# Individual differences

“Weak chunking”

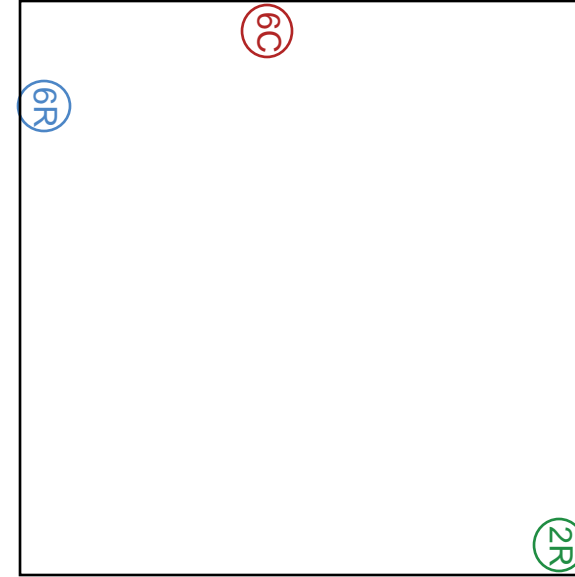
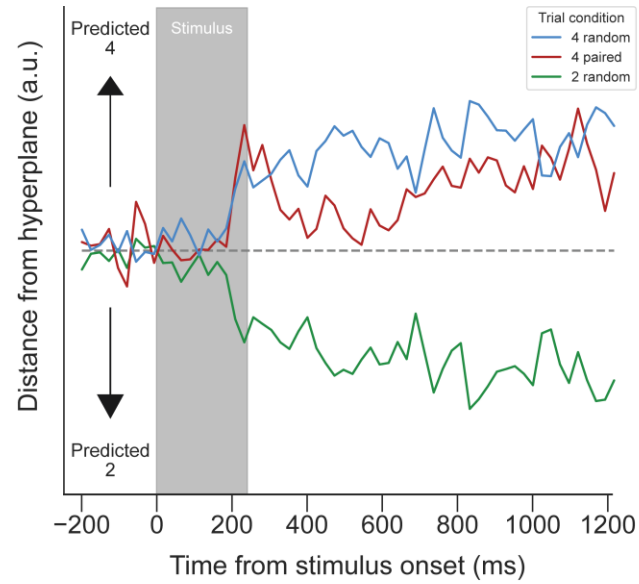


“Strong chunking”

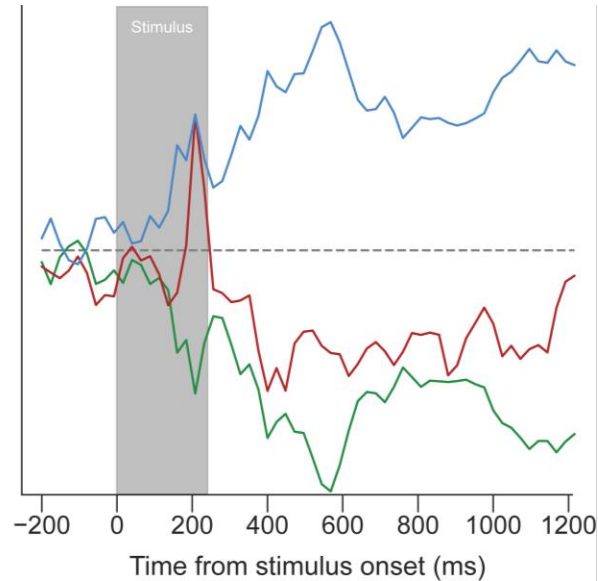


# Individual differences

“Weak chunking”

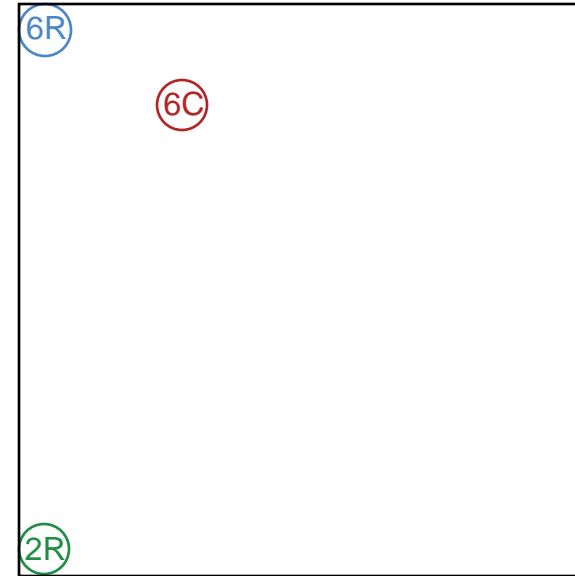
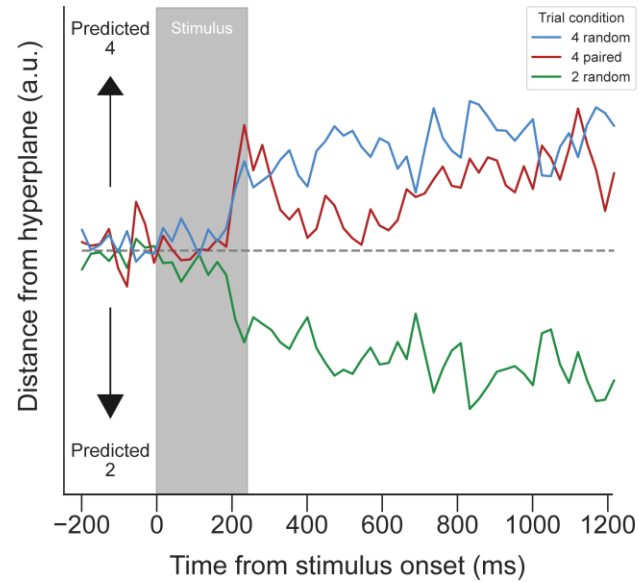


“Strong chunking”

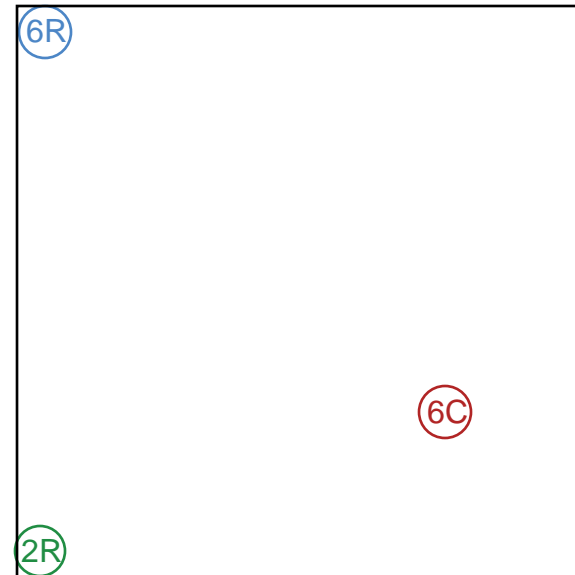
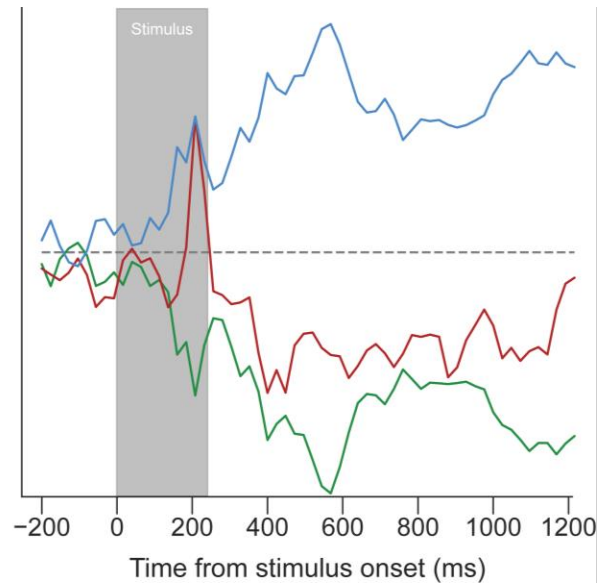


# Individual differences

“Weak chunking”

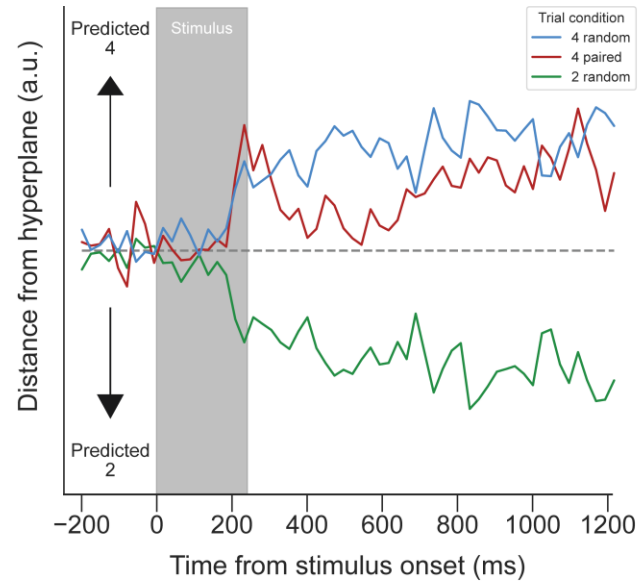


“Strong chunking”

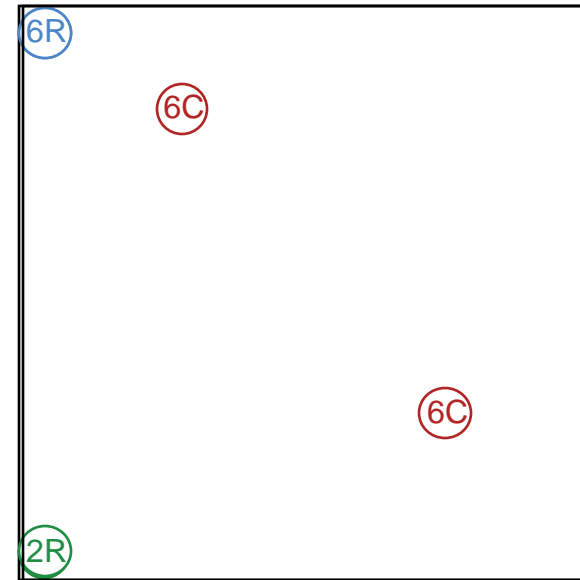
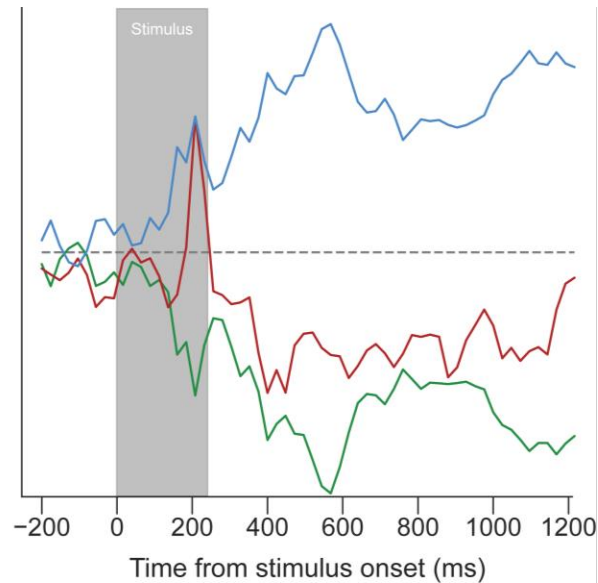


# Individual differences

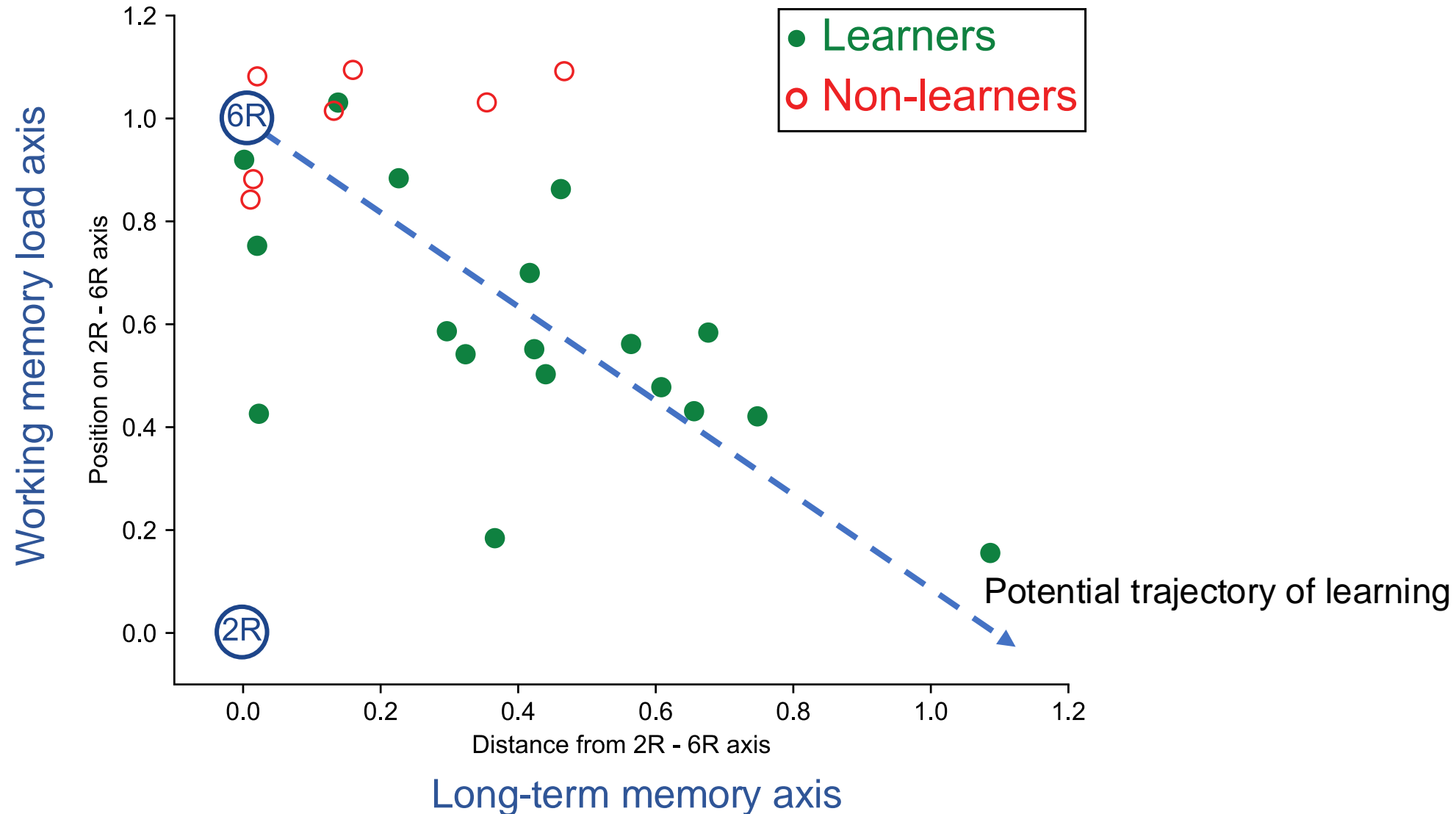
“Weak chunking”



“Strong chunking”



# Multidimensional scaling on each subject



# Conclusions

- A multivariate neural signal for items in working memory shows associative learning *reduces* the number of items stored in working memory
- Furthermore, neural signatures of associative learning showed the reduction only in those that **successfully learnt the associations**
- This is consistent with a *chunking* account – associative learning may not allow one to circumvent item pointer limits