

# Associative learning changes multivariate neural signatures of visual working memory

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Henry Jones



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William Thyer

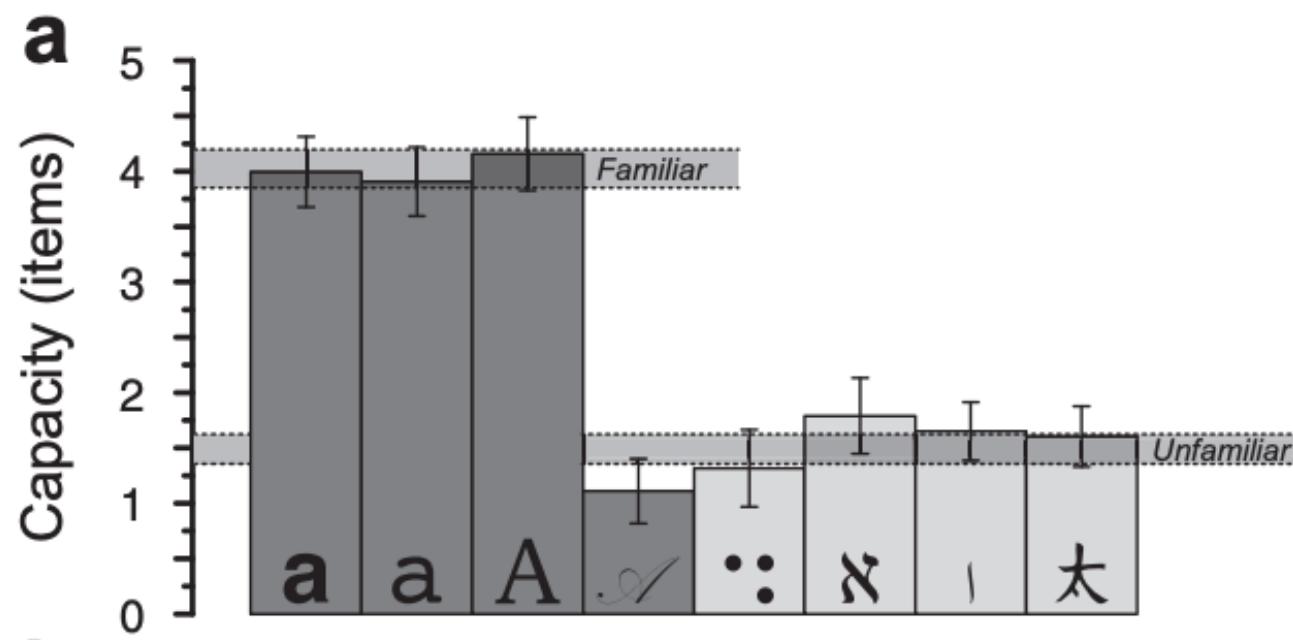


Edward Awh

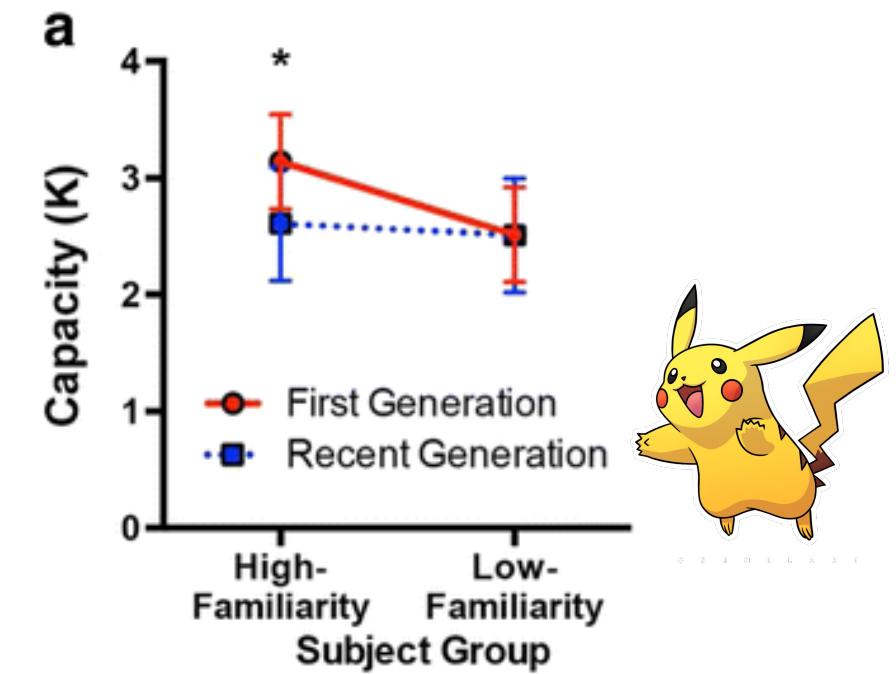


# Working memory is aided by long-term memory

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

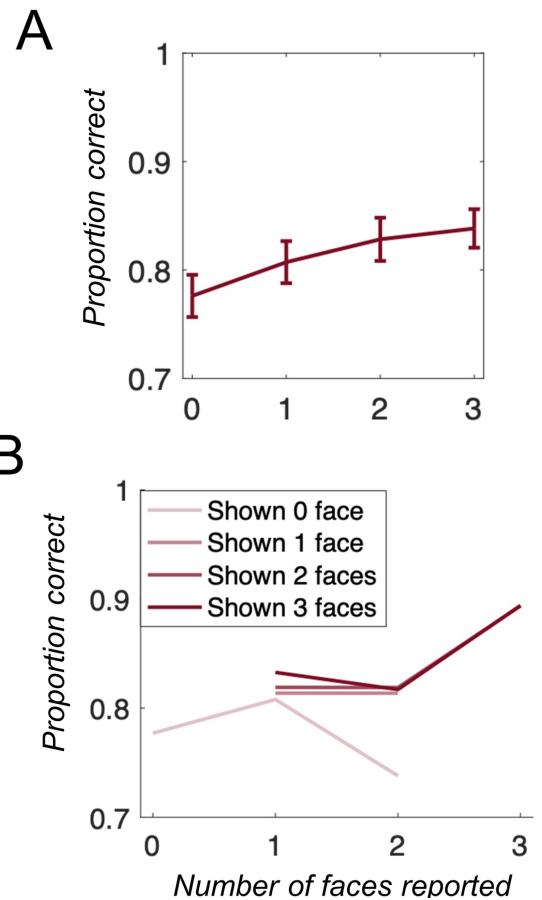
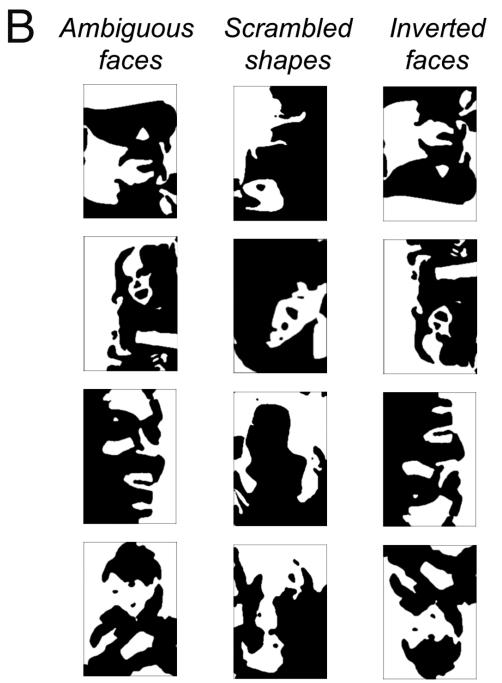


Ngiam et al. (2019) *JEP:G*

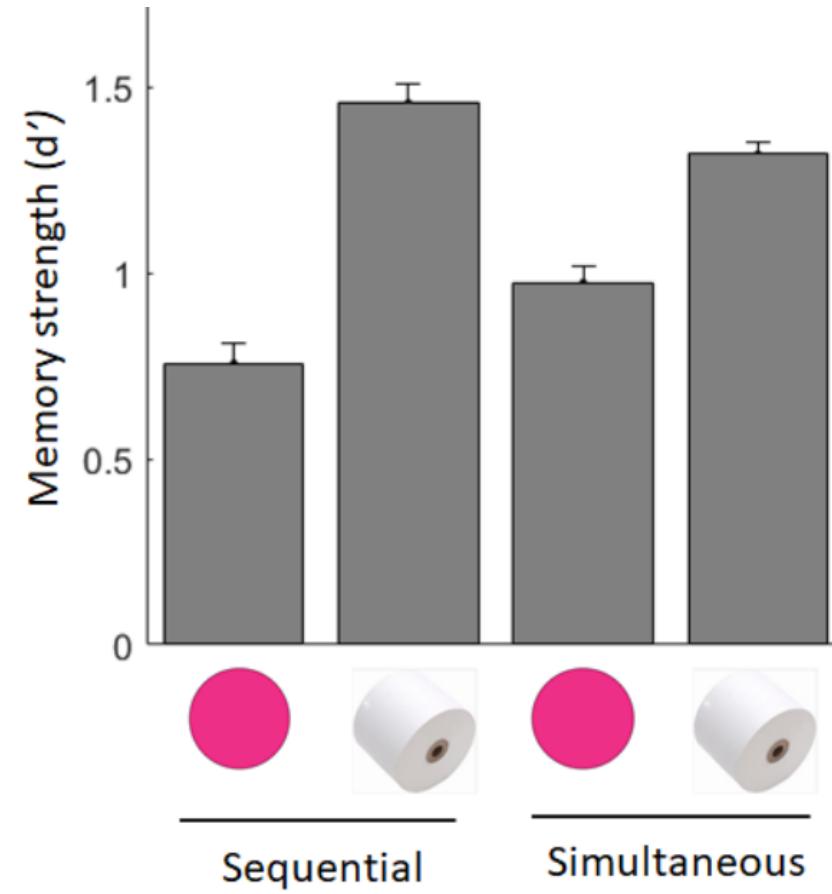


Xie and Zhang (2017) *M&C*

- Meaningfulness // real-world objects:

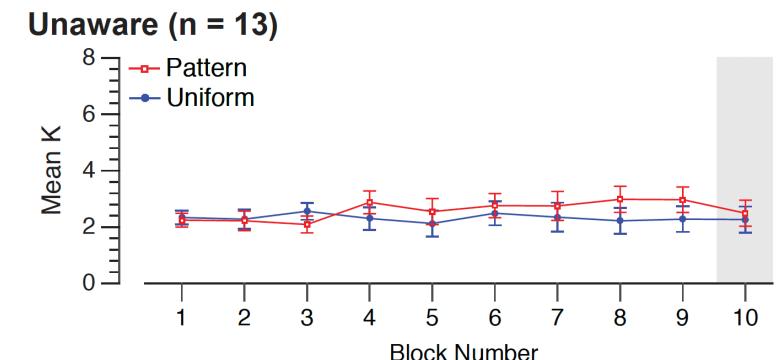
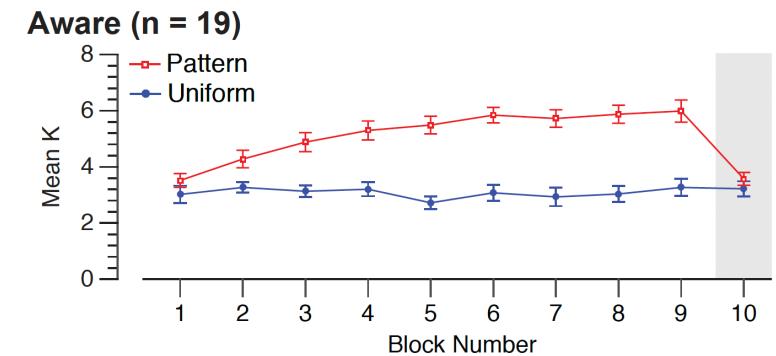
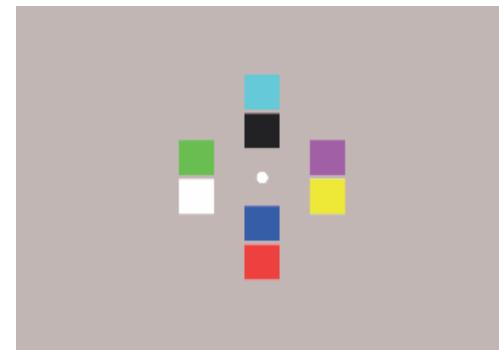
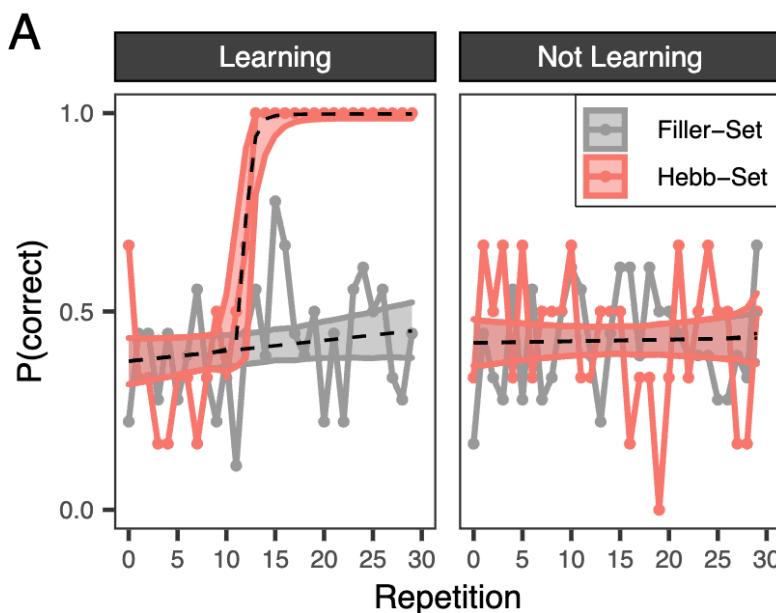
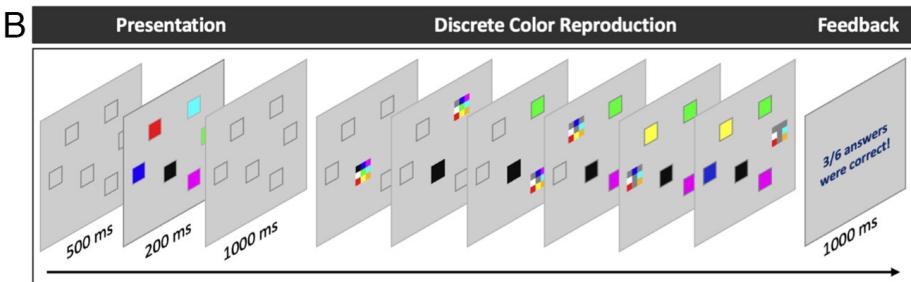
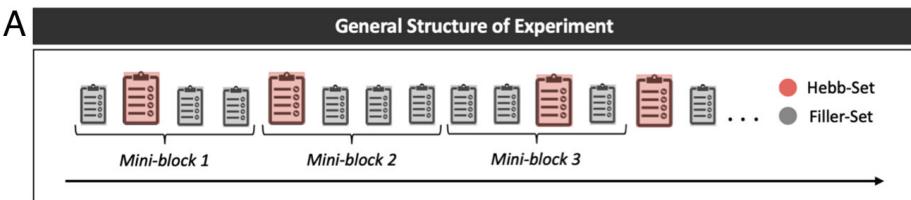


Asp et al. (2021) *JoCN*



Brady and Störmer (2022) *JEP:LMC*

# • Repetition learning // associative learning:

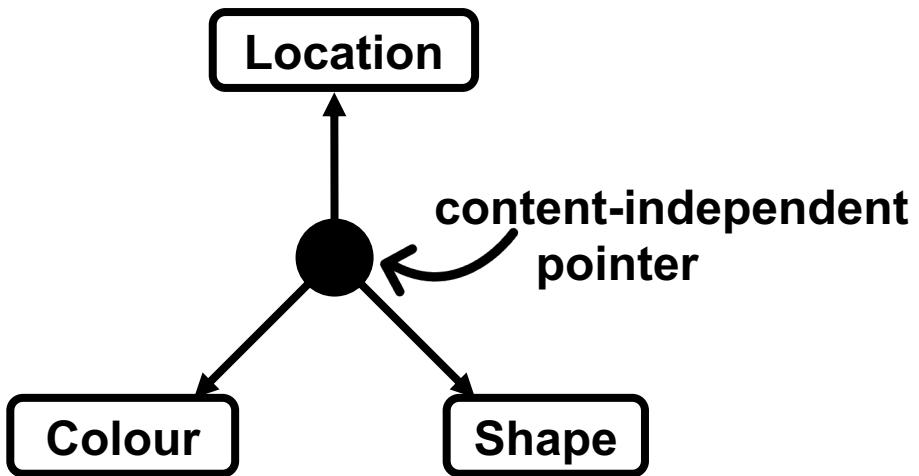


# How is working memory performance improved?

- Working memory load may be ***expanded*** for familiar // meaningful // learned stimuli
  - Additional resources are recruited allowing a greater number of items to be held within working memory
- Working memory load is reduced via ***chunking***
  - Load is reduced by requiring fewer “chunks” to be held in working memory
  - Recall is improved by relying on recruitment of long-term memory

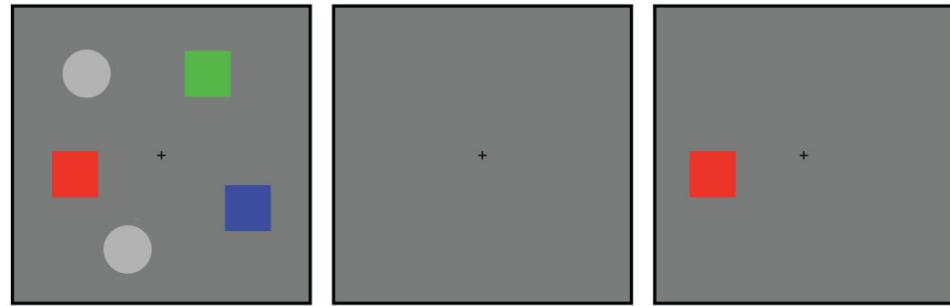
# Pointers in working memory

- Pylyshyn (2009) proposed the visual system has an indexing mechanism that keeps track of an individual object through its changes
  - This index is **abstracted** from the contents of the object
- We propose that **items** in working memory are assigned to a **content-independent pointer**

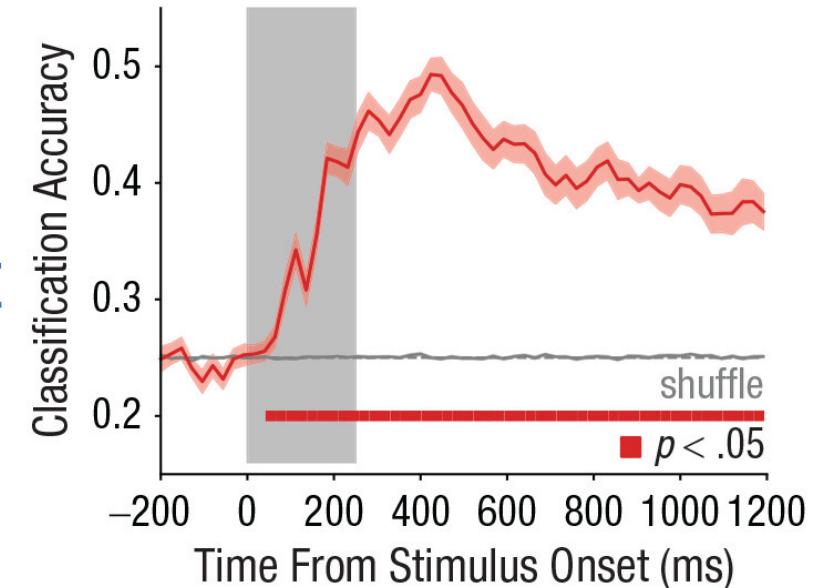


# Multivariate neural signature of WM pointers

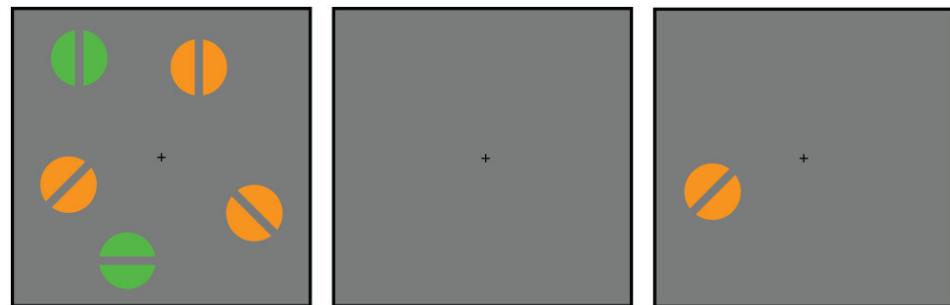
Experiment 1: Color



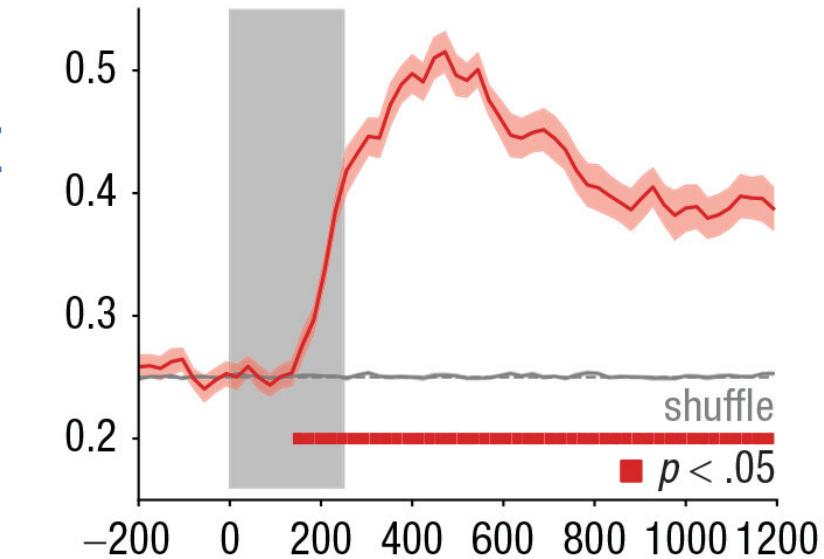
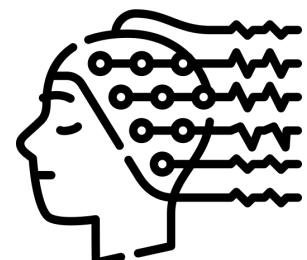
Train and test



Experiment 2: Orientation

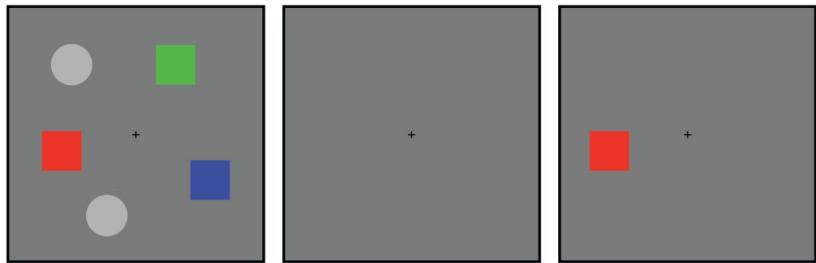


Train and test



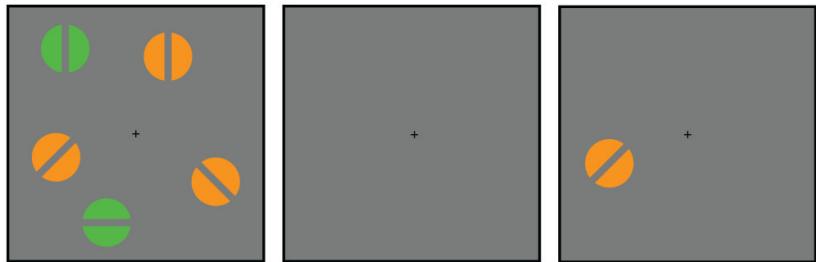
# Multivariate neural signature of WM pointers

Experiment 1: Color



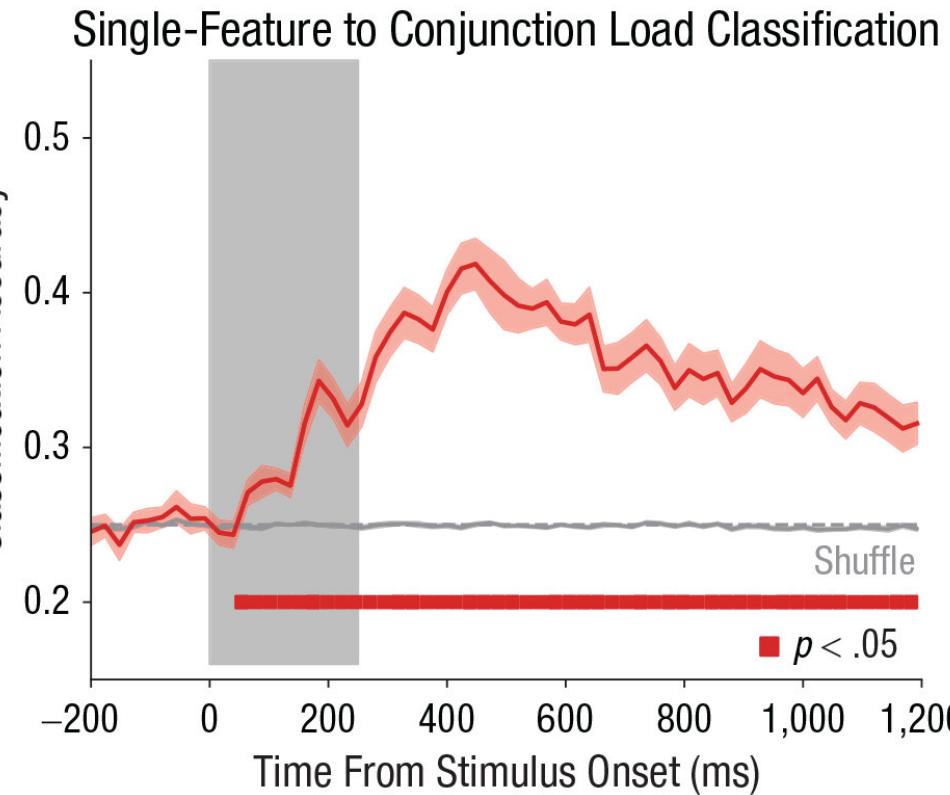
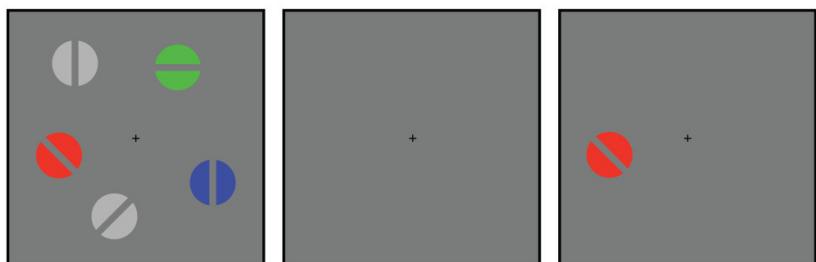
Train

Experiment 2: Orientation



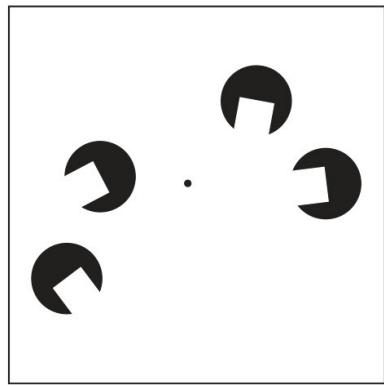
Test

Experiment 3: Conjunction

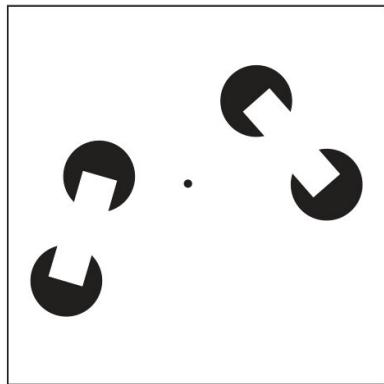


# Multivariate neural signature of WM pointers

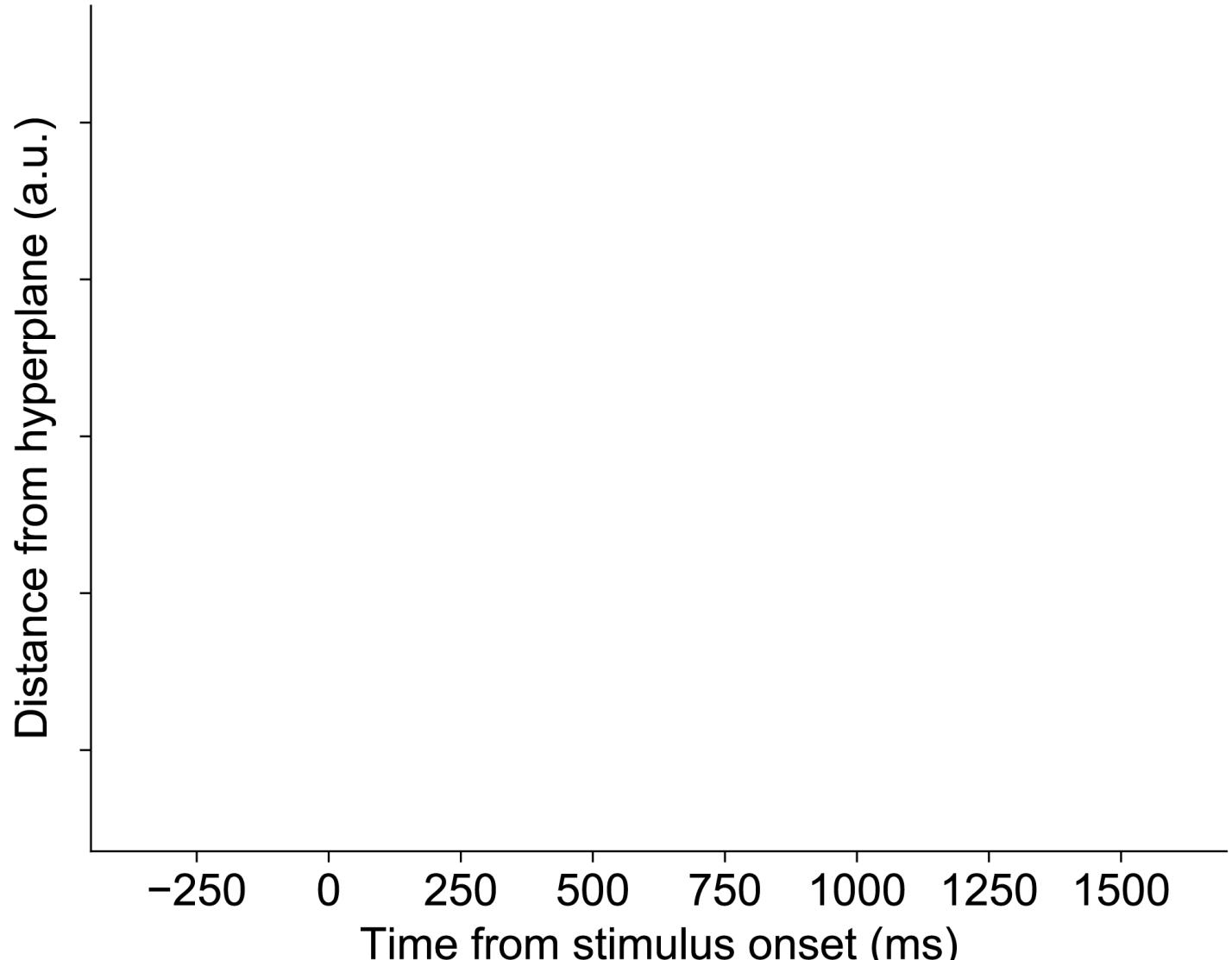
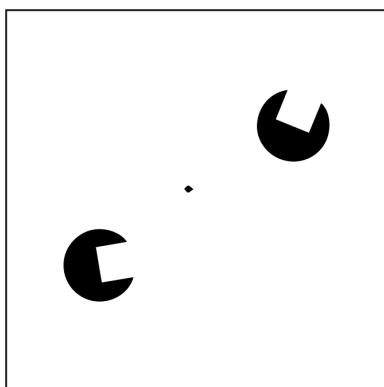
4 Ungrouped



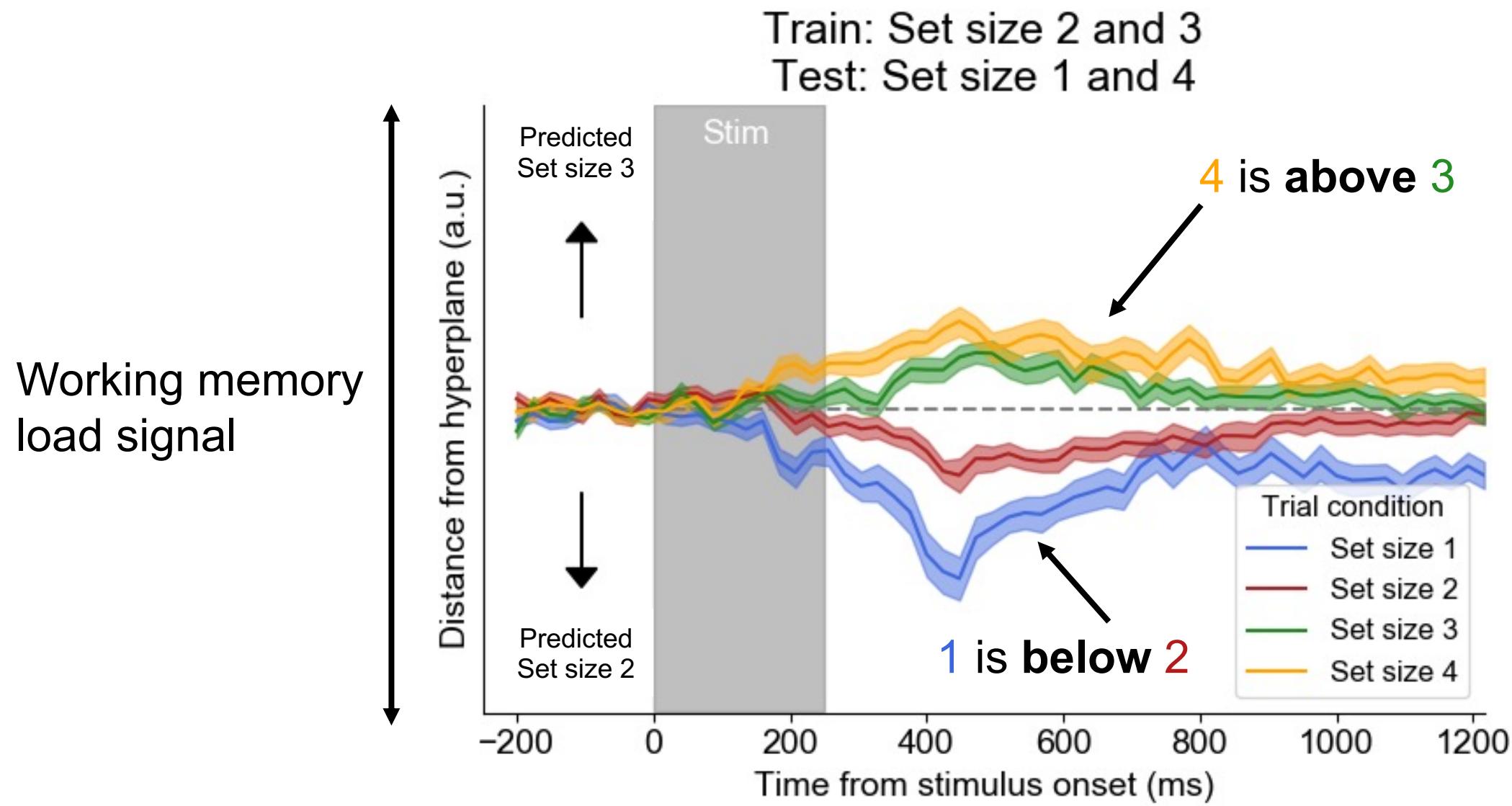
4 Grouped



2 Ungrouped



# Different working memory loads on the hyperplane



# Multivariate neural signature of WM pointers



Henry Jones

- The multivariate load signal for pointers is dissociated from spatial attention *Jones et al. (accepted), Psychological Science*
- The load signal generalizes from color to motion coherence of random dot kinematograms *Henry's VSS talk this year*



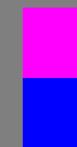
Darius Suplica

- The multivariate load signal is shared for audio and visual stimuli *in prep*

How do working memory pointers change with associative learning?

# 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?

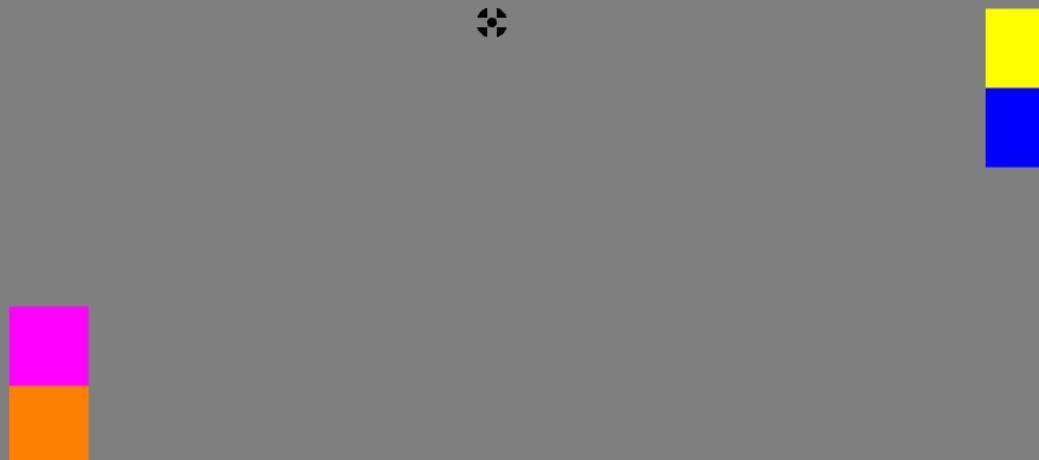


■ . ■



# Experiment 1: Pre-training and post-training

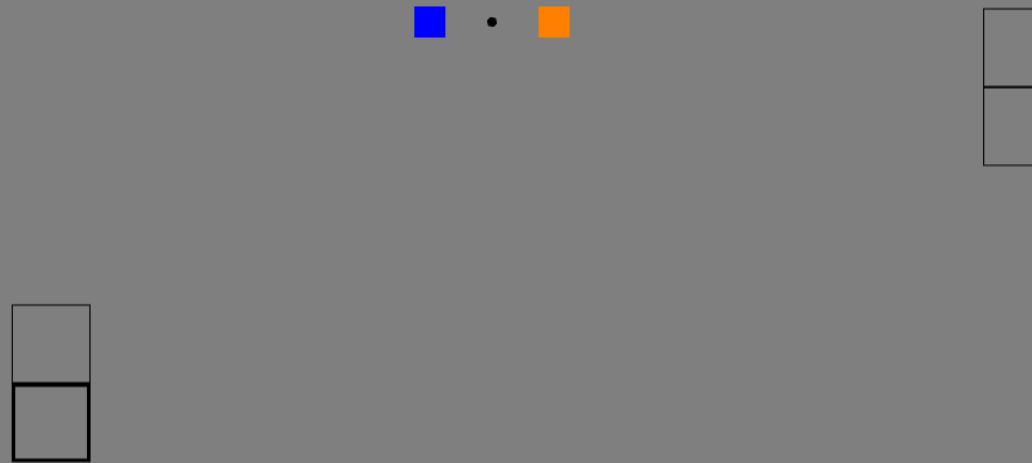
- Before training – 4 random colors
- After training – 4 paired colors (two learned pairs)



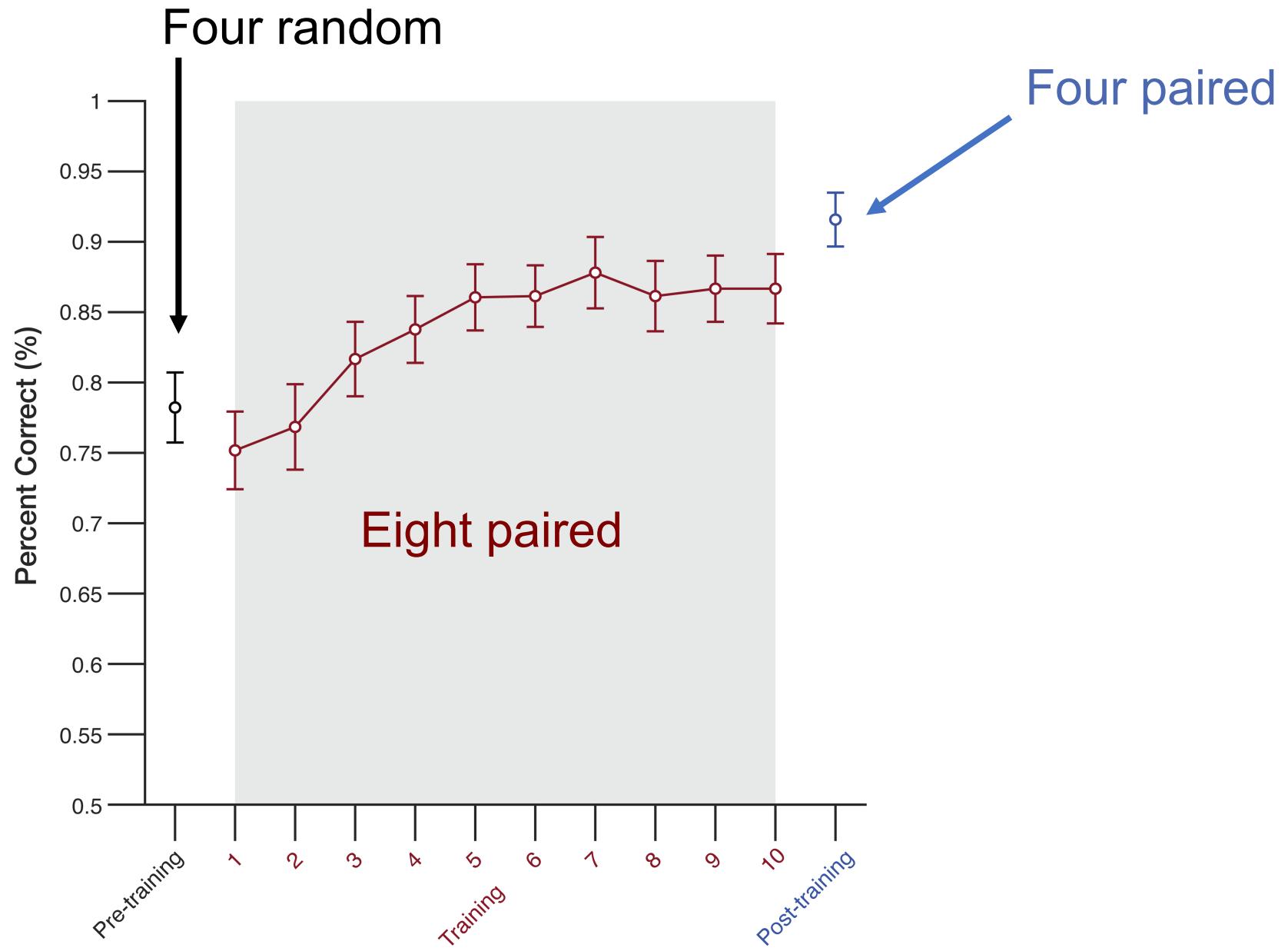
# Experiment 1: Pre-training and post-training



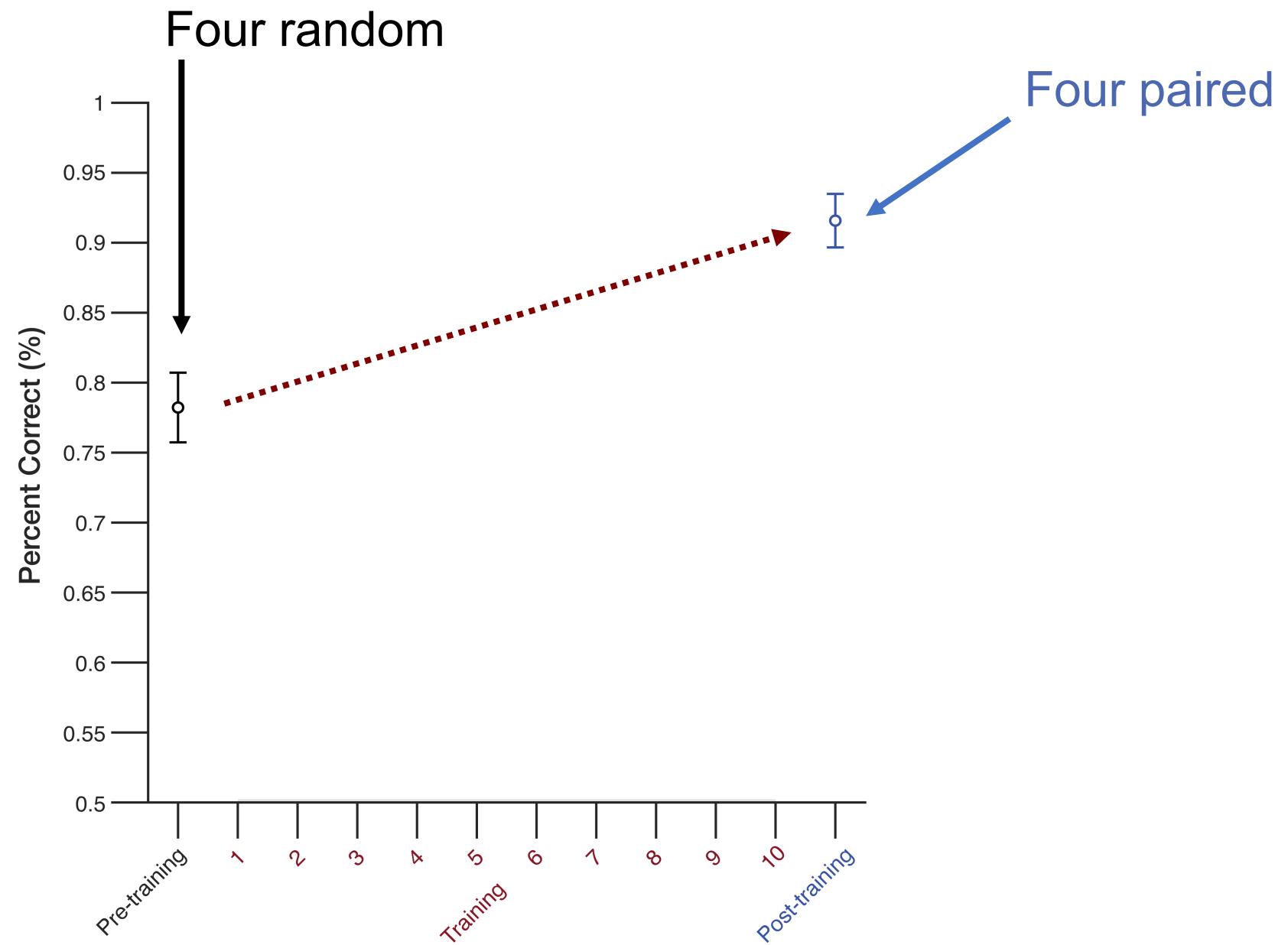
# Experiment 1: Pre-training and post-training



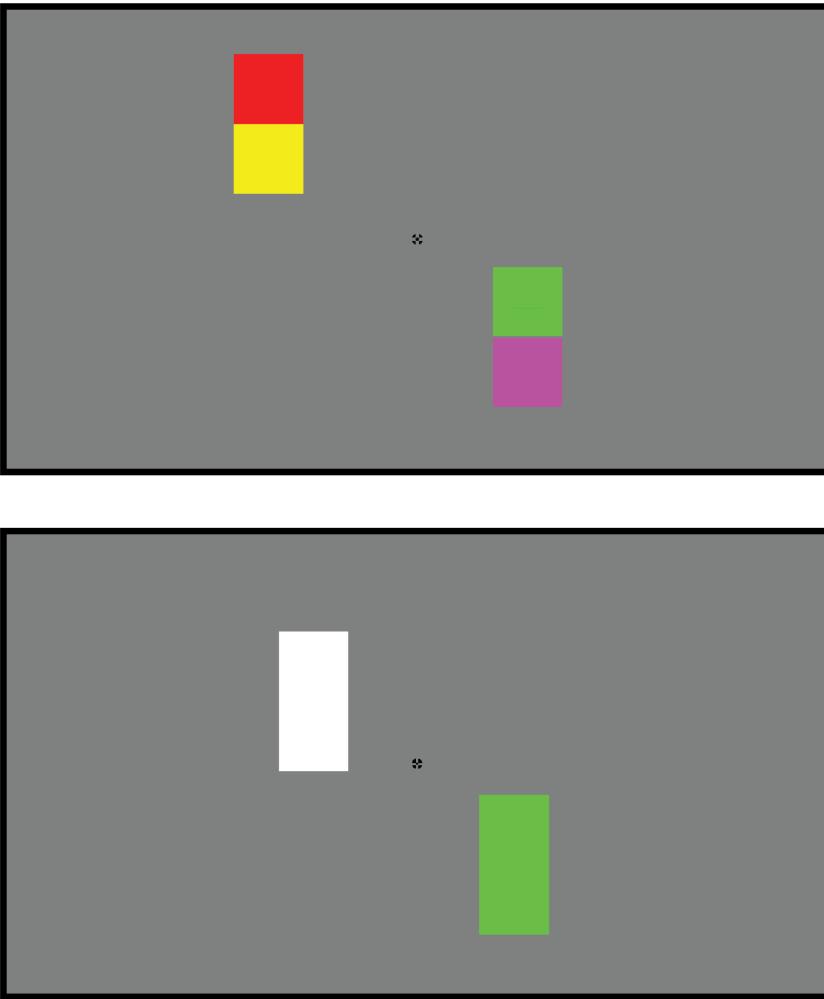
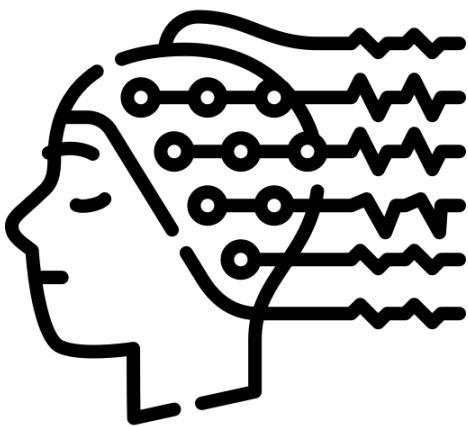
# E1: training session – aggregate performance



# E1: training session – average performance



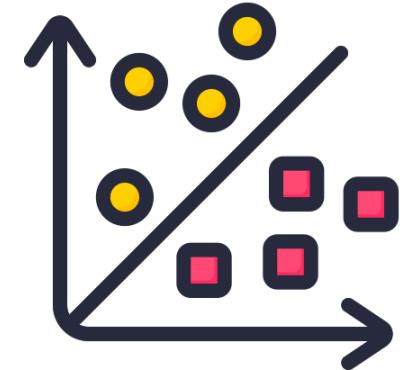
# Experiment 1: EEG session



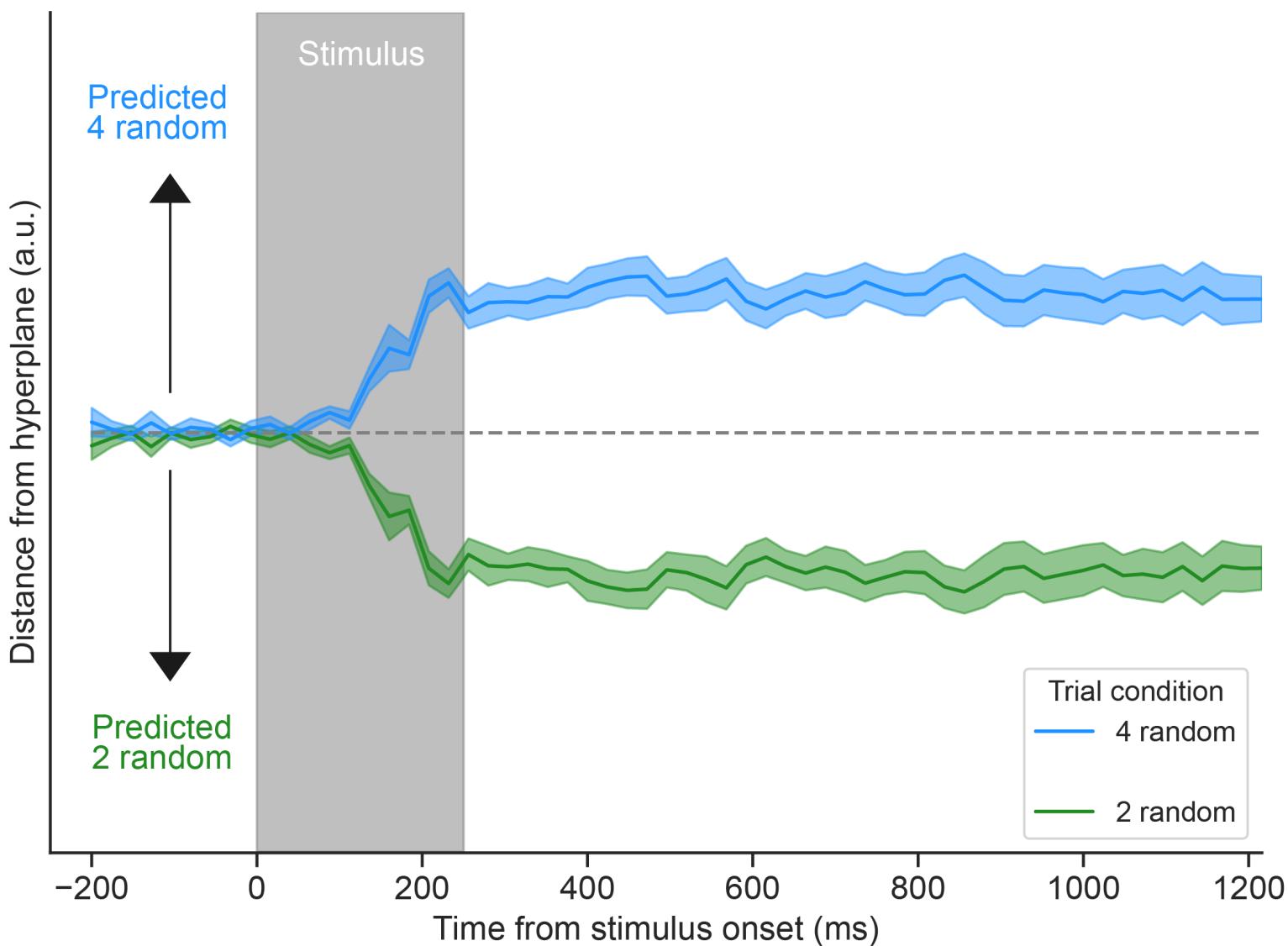
Four random  
Four paired

Perceptually equivalent

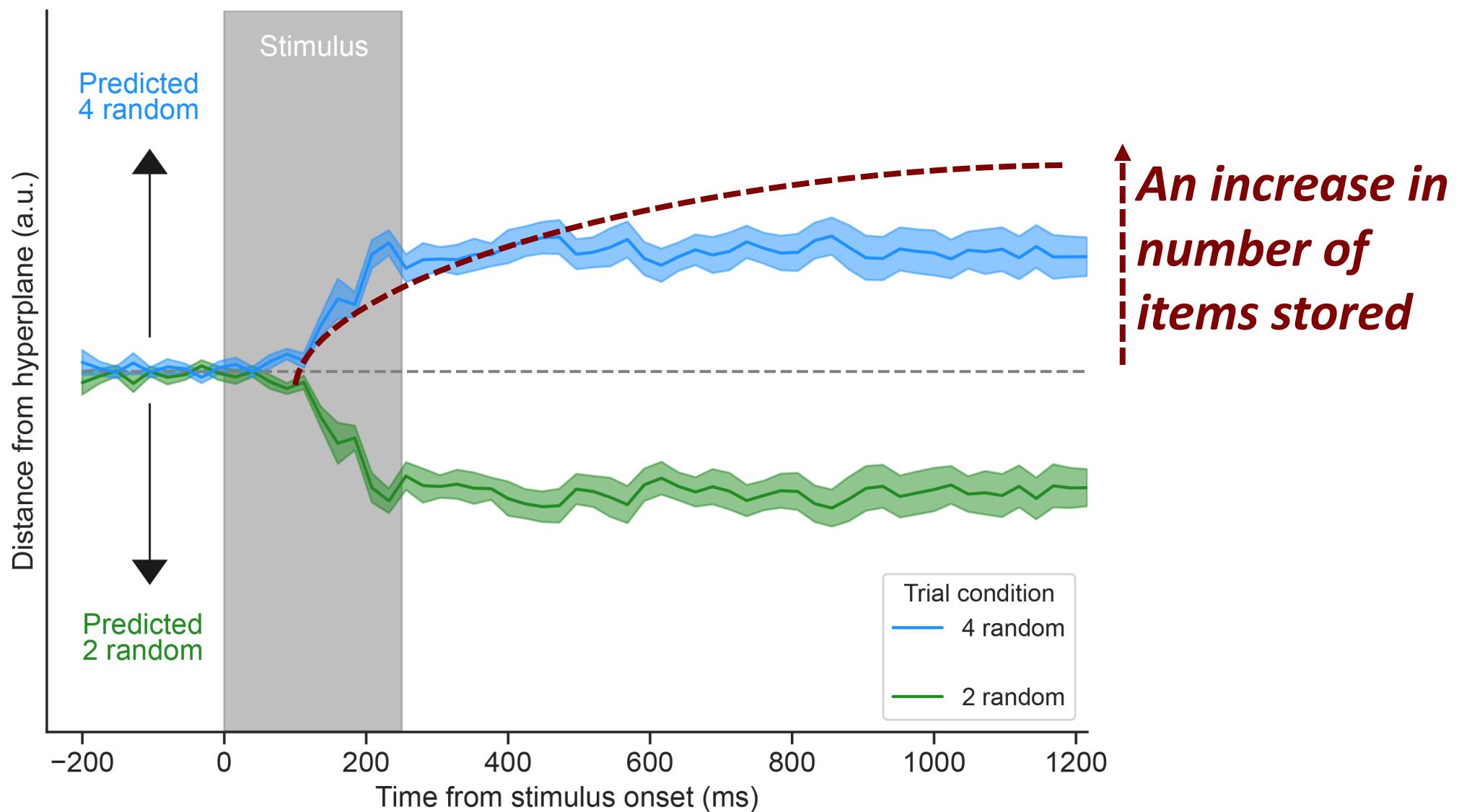
Two random



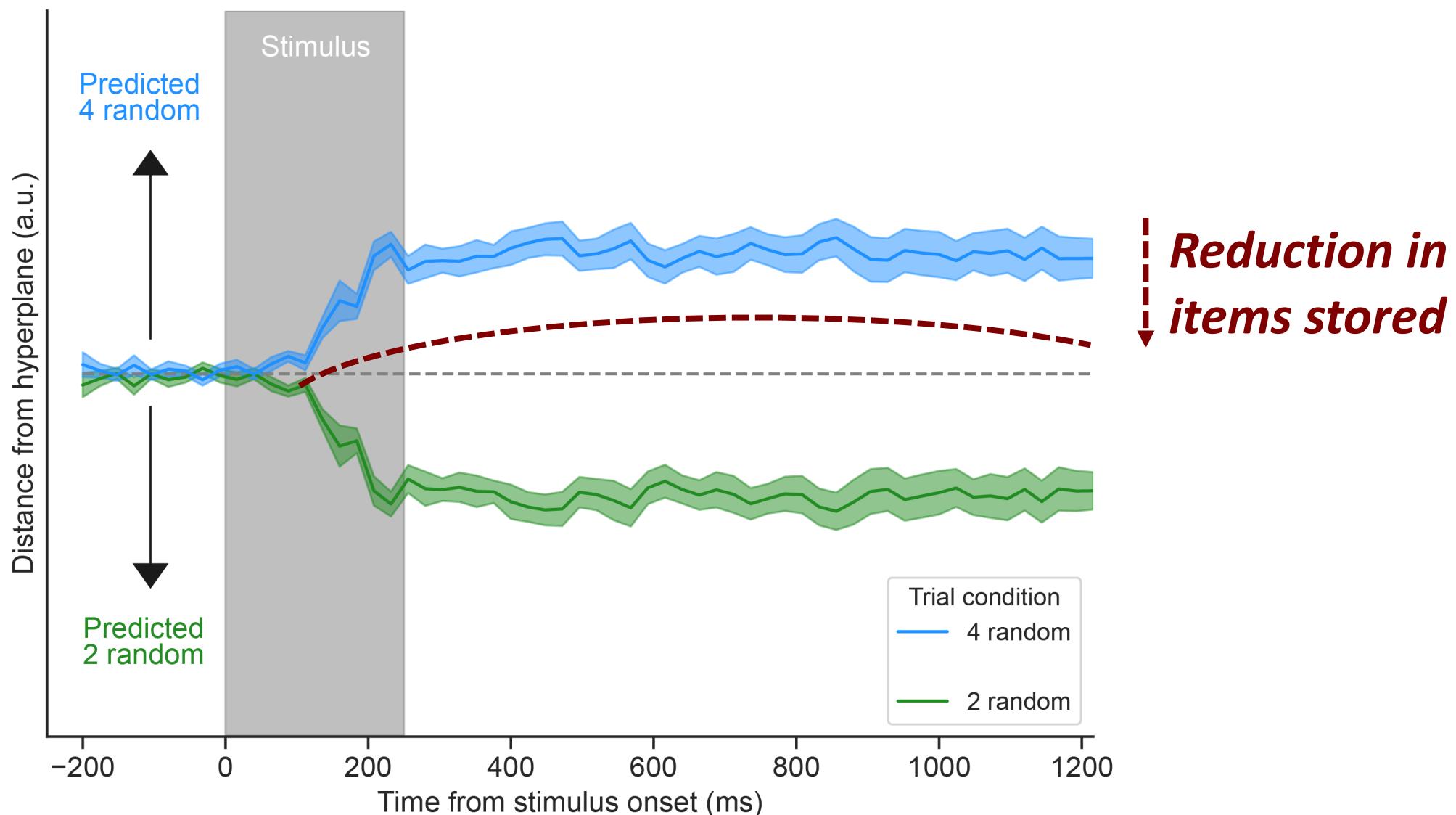
# Train 2 random versus 4 random



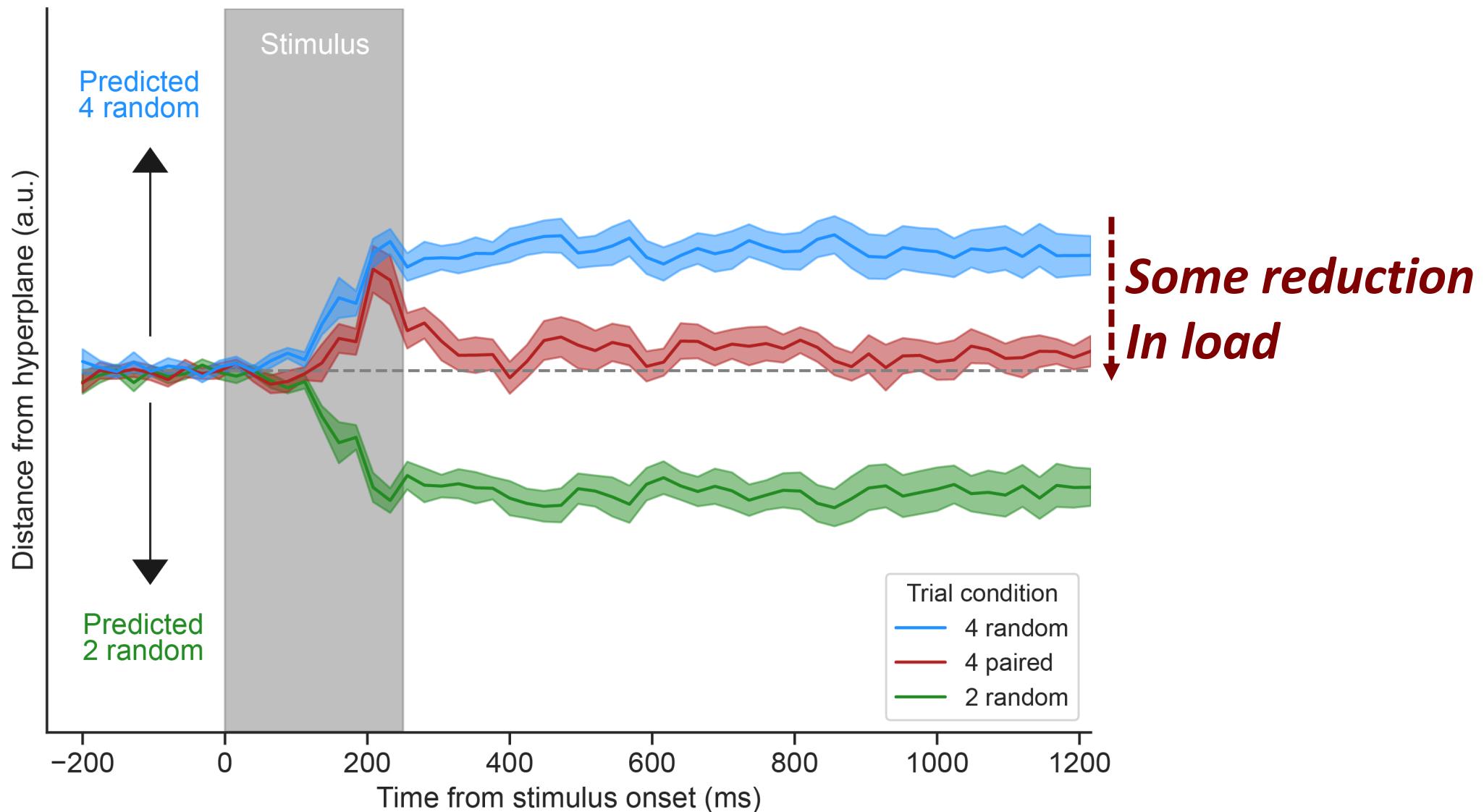
# Train 2 random versus 4 random, where is 4 paired?



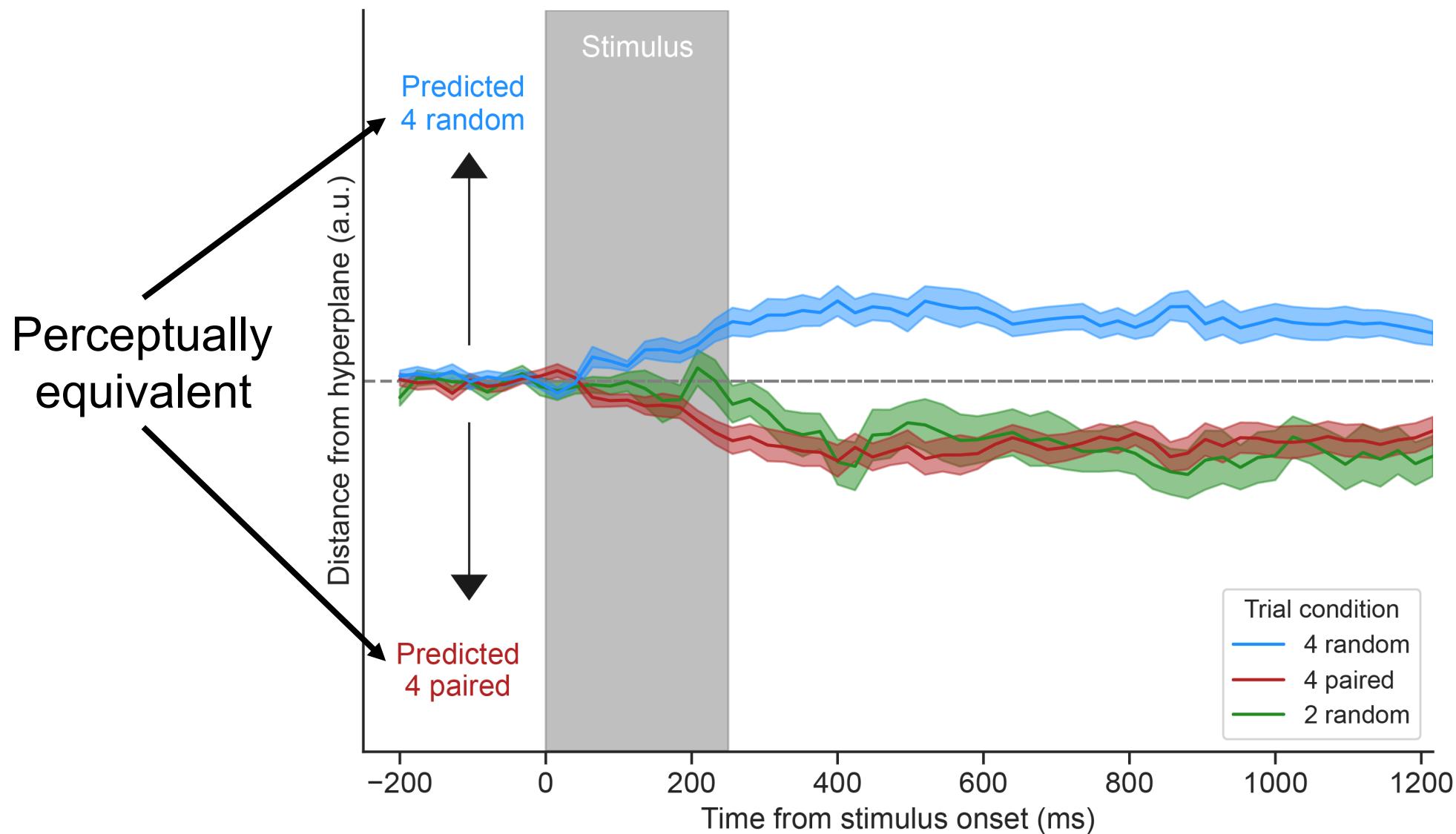
# Train 2 random versus 4 random, where is 4 paired?



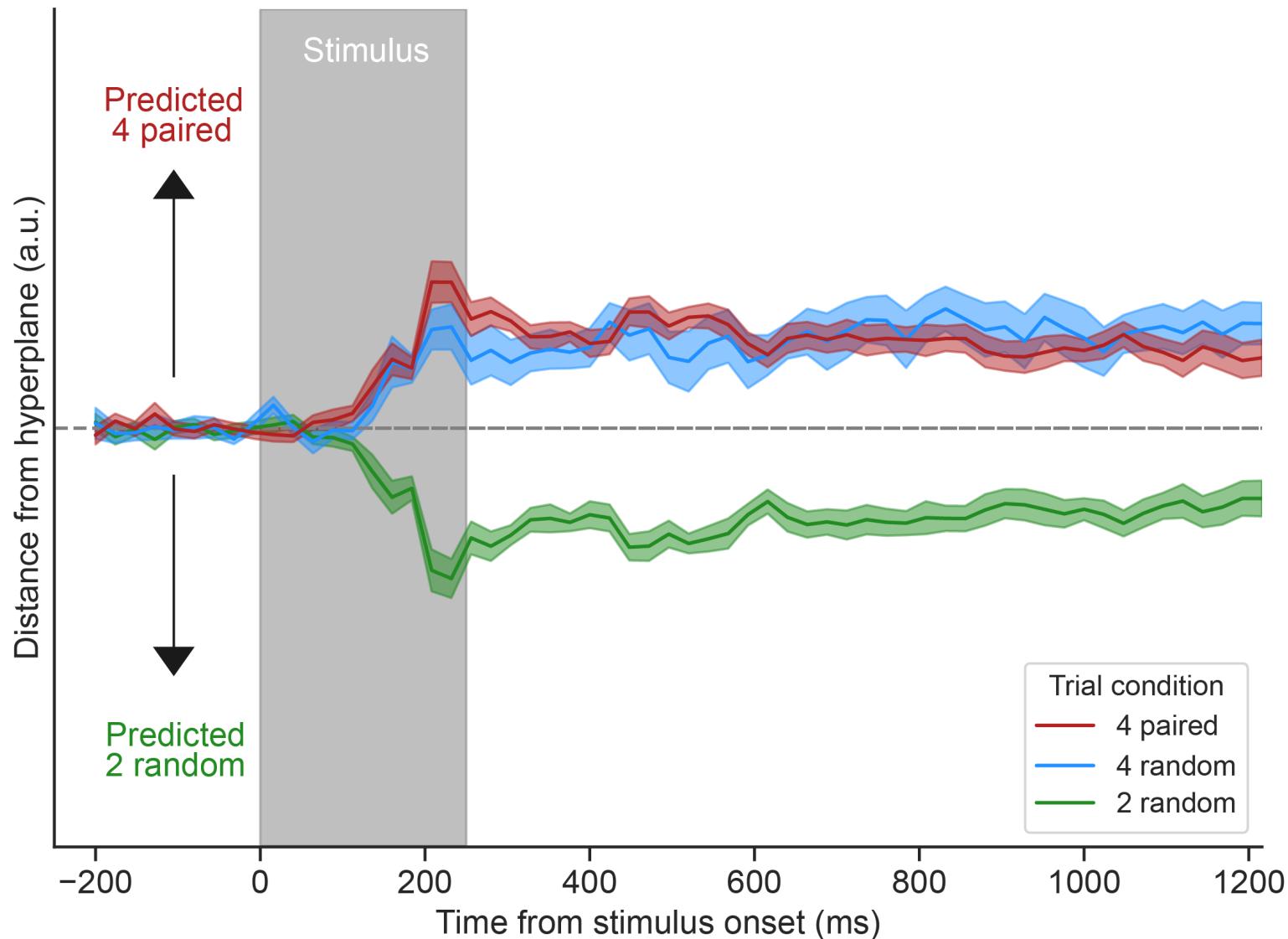
# Train 2 random versus 4 random, test 4 paired



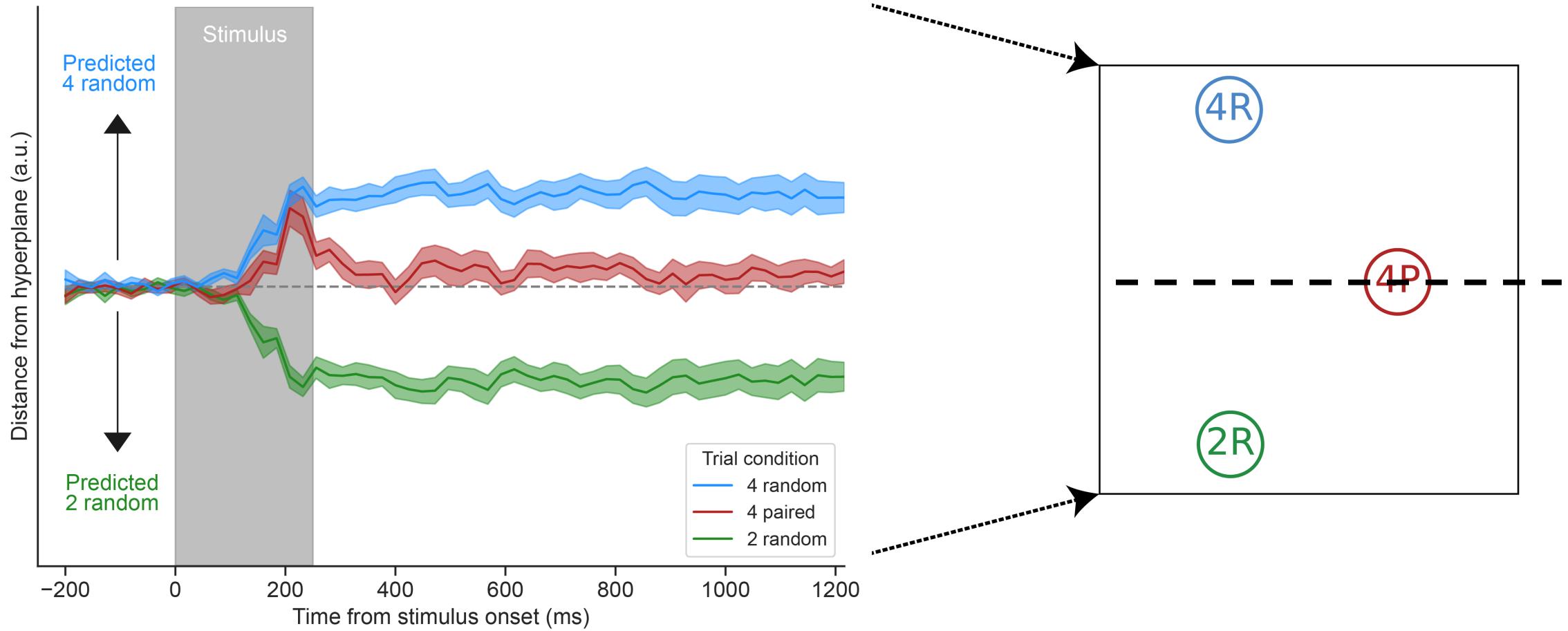
# Train 4 random versus 4 paired, test 2 random



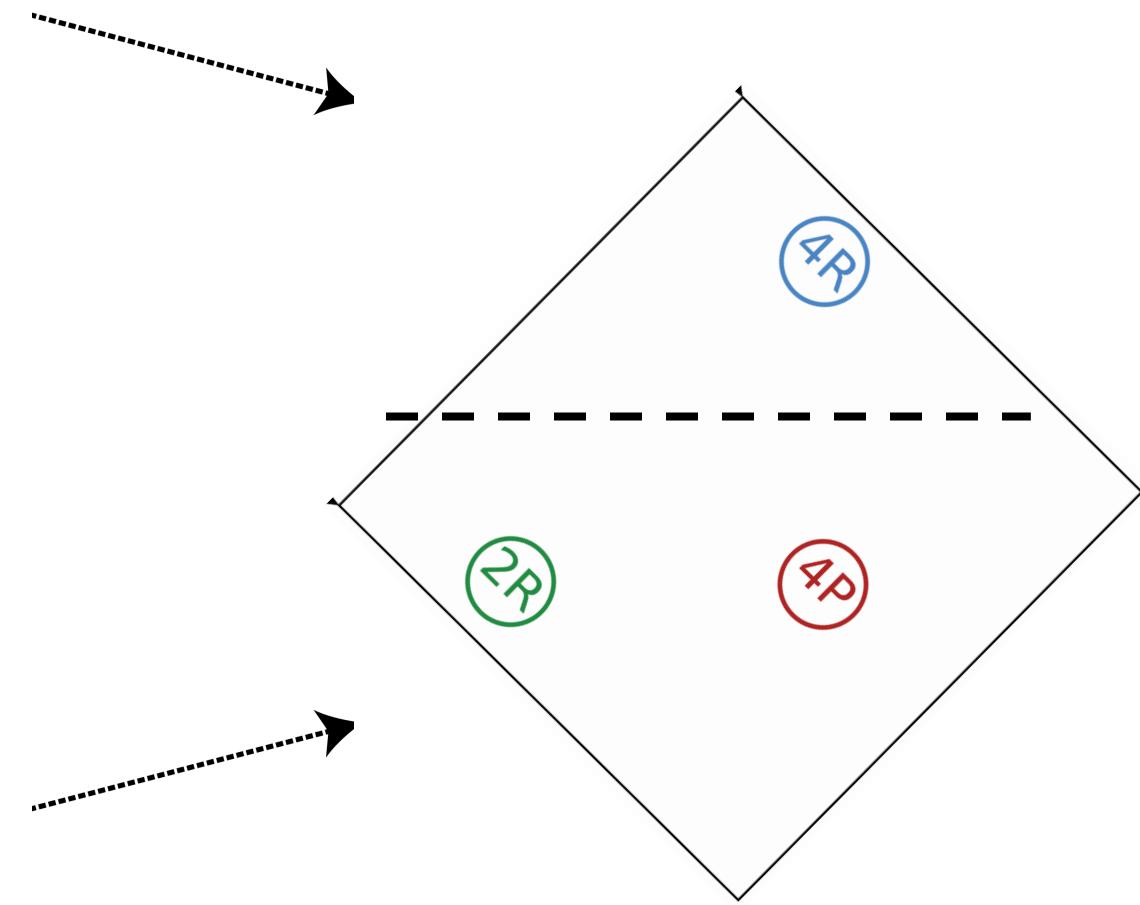
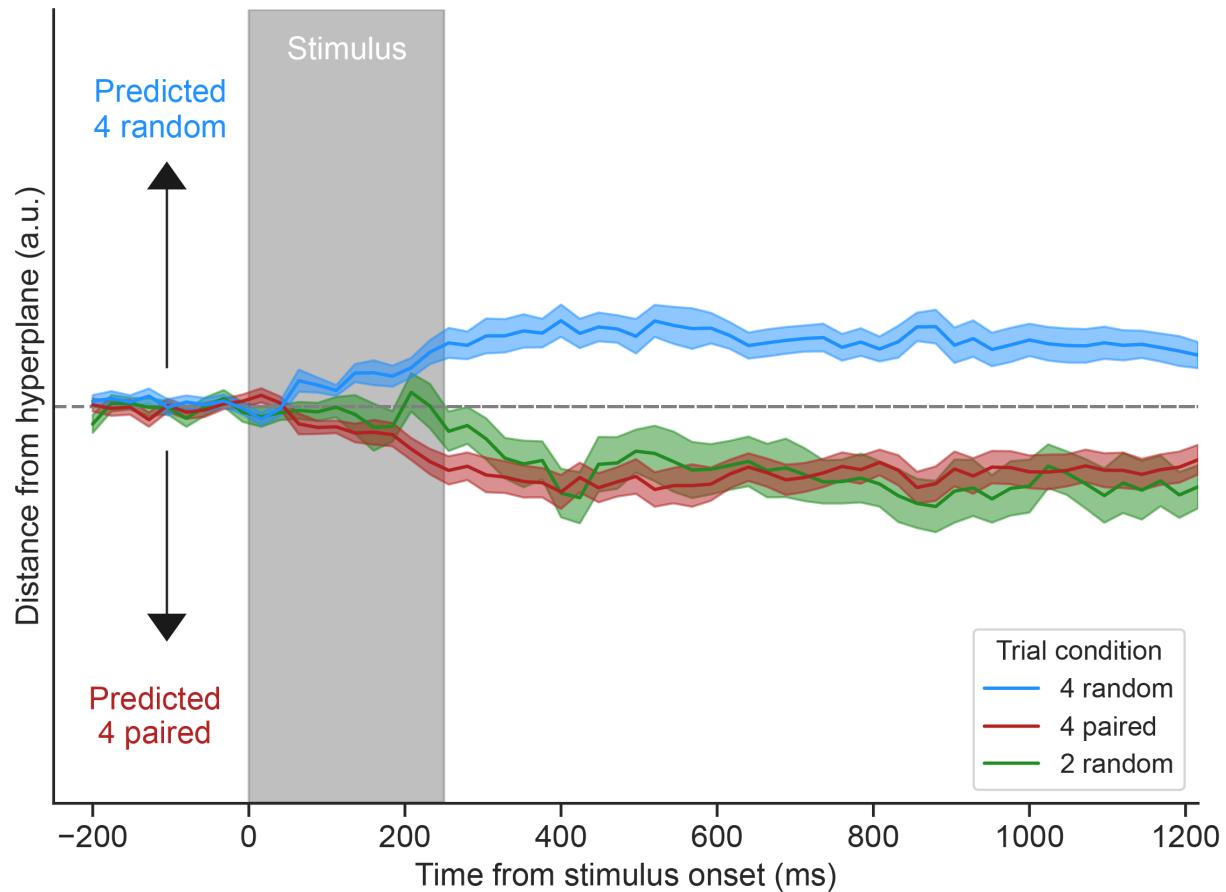
# Train 2 random versus 4 paired, test 4 random



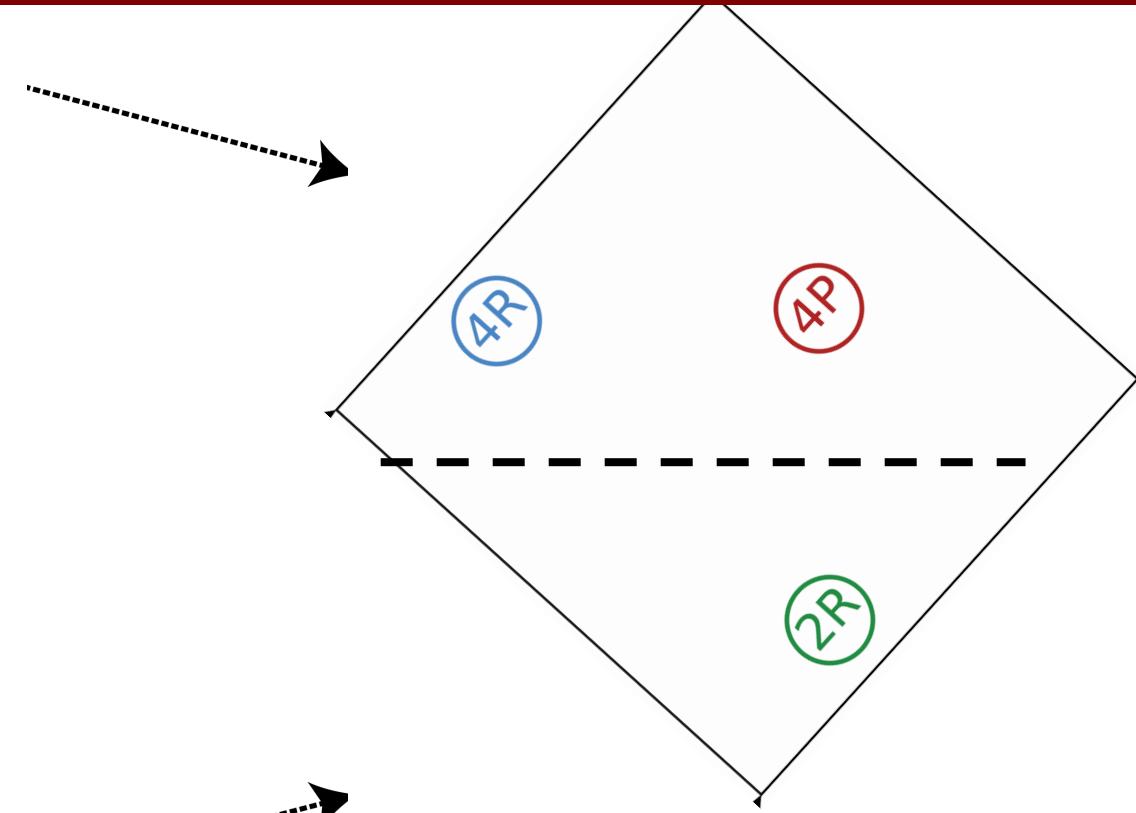
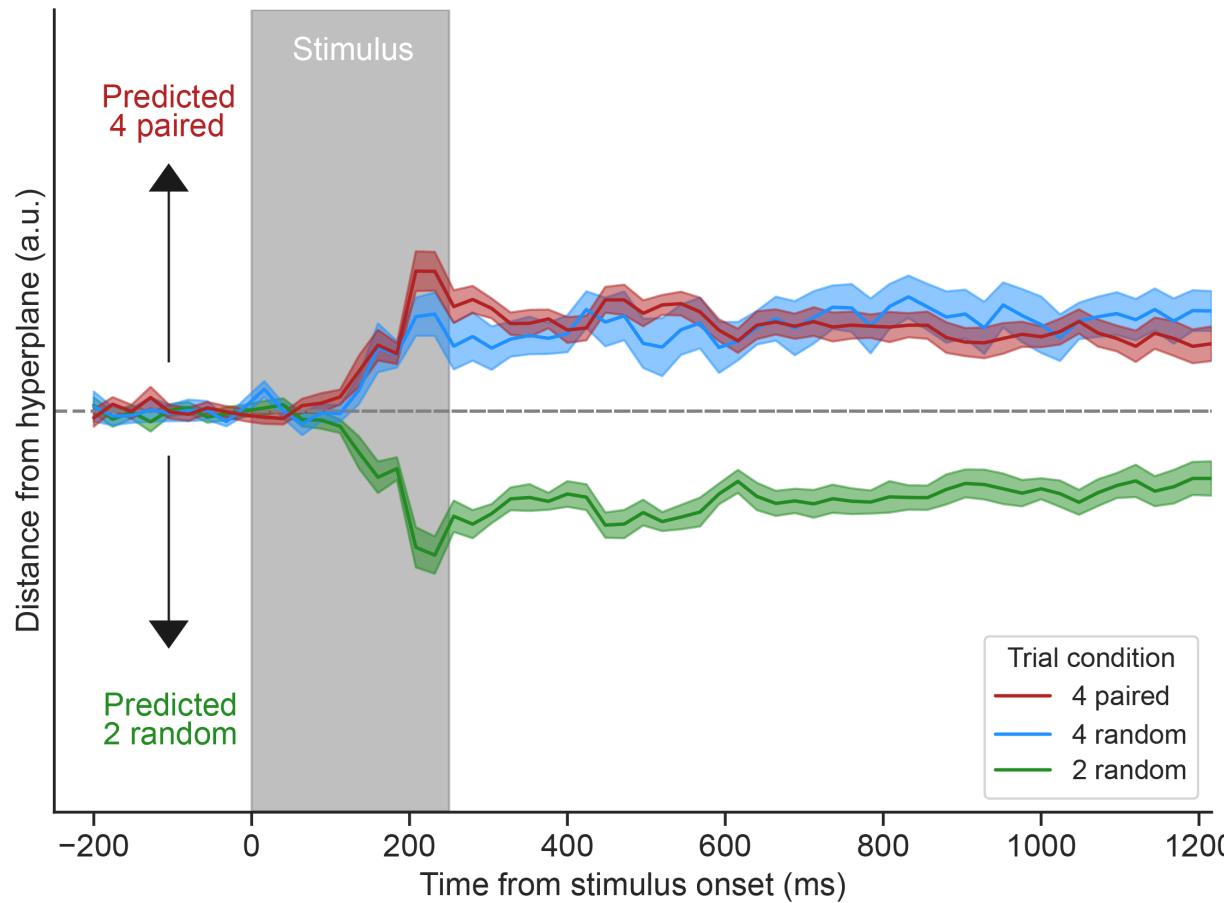
# Multidimensional scaling



# Multidimensional scaling

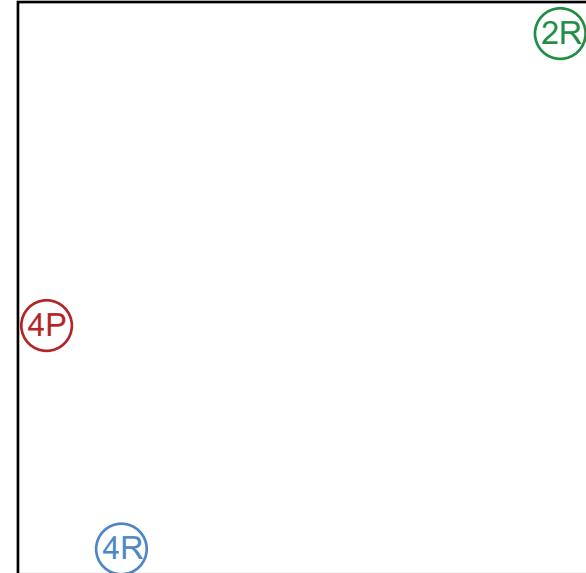
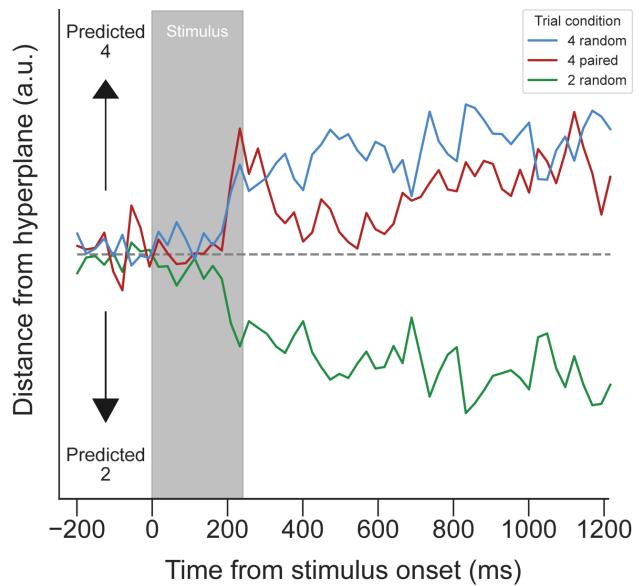


# Multidimensional scaling

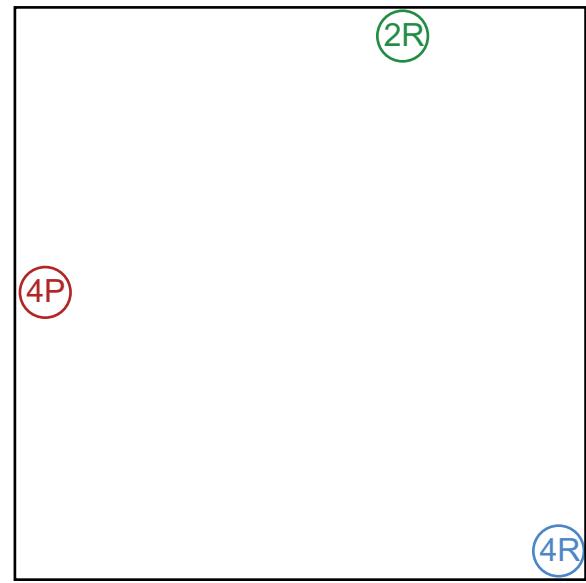
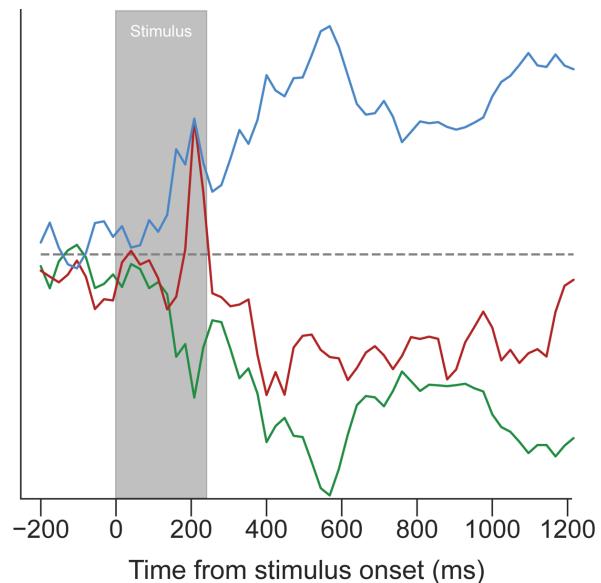


# Individual differences

“Weak chunking”

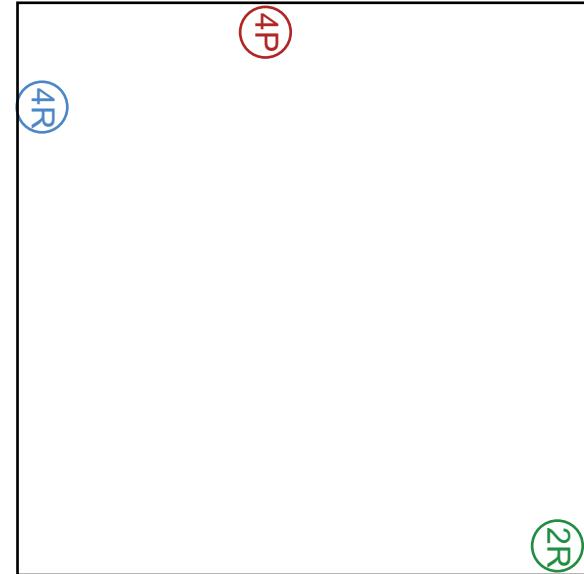
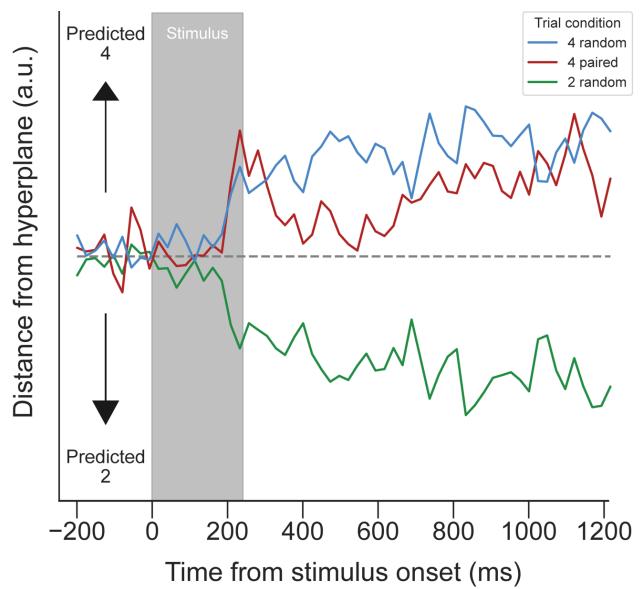


“Strong chunking”

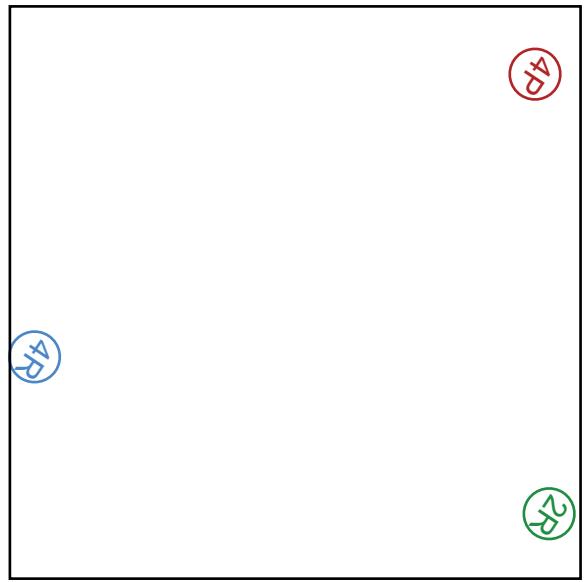
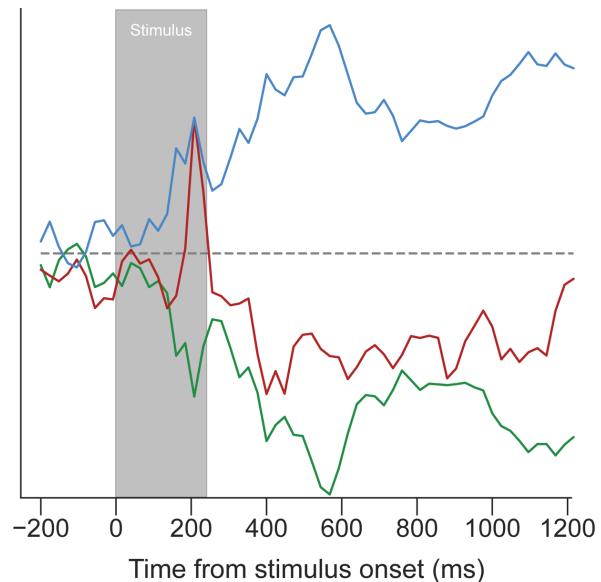


# Individual differences

“Weak chunking”

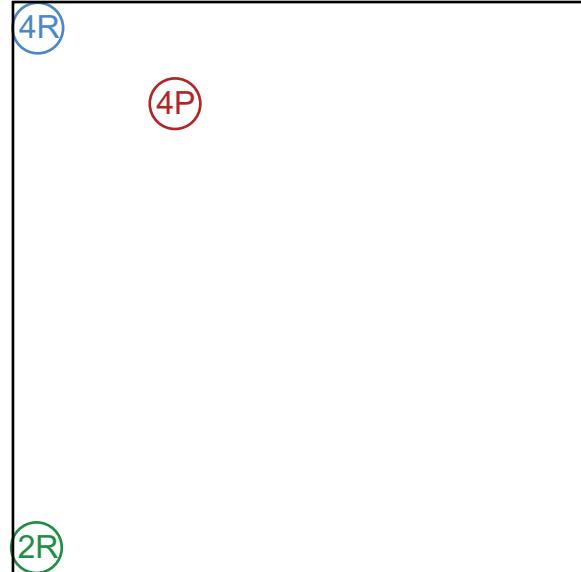
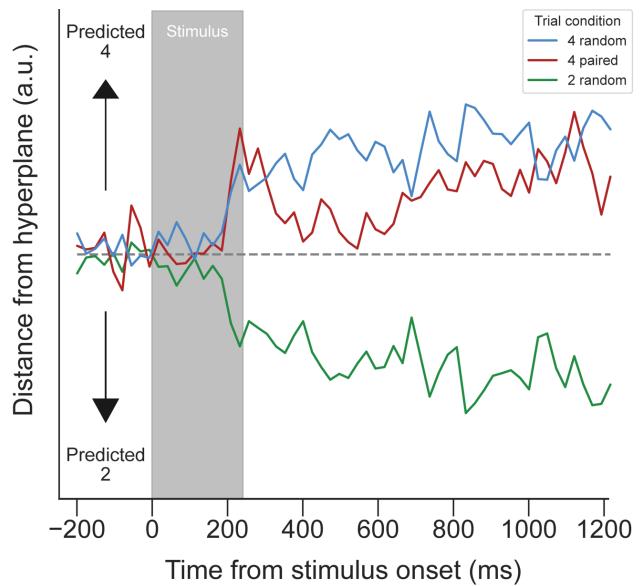


“Strong chunking”

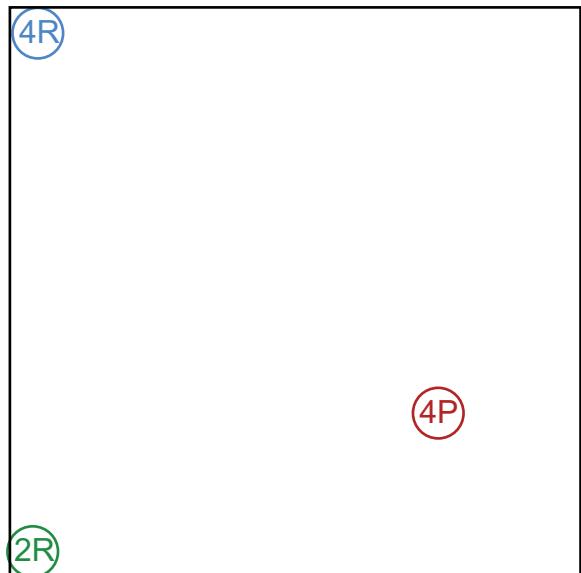
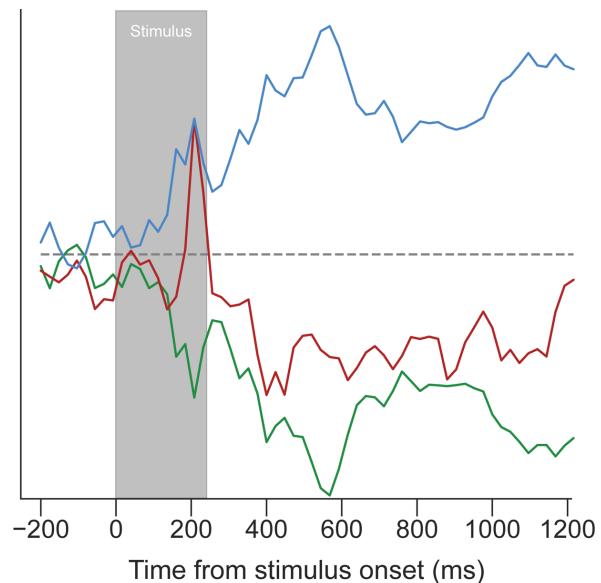


# Individual differences

“Weak chunking”

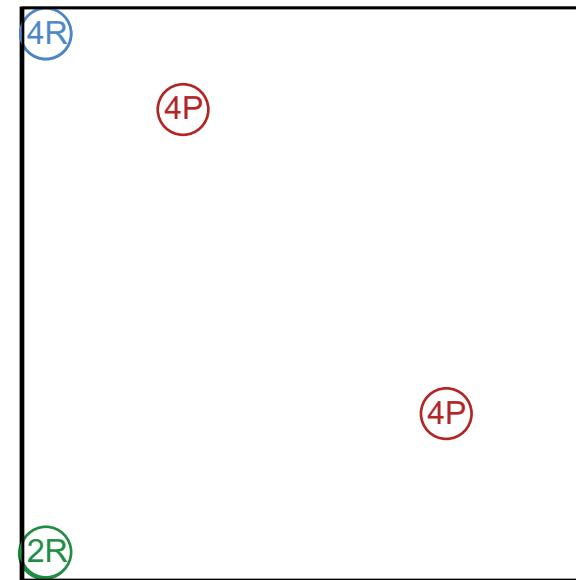
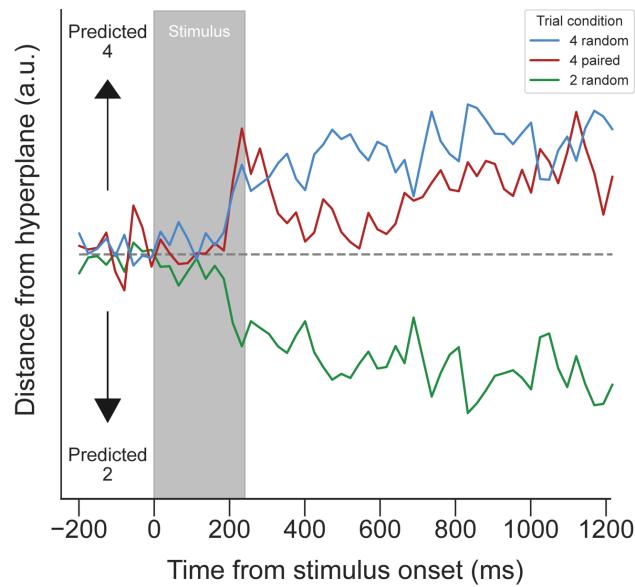


“Strong chunking”

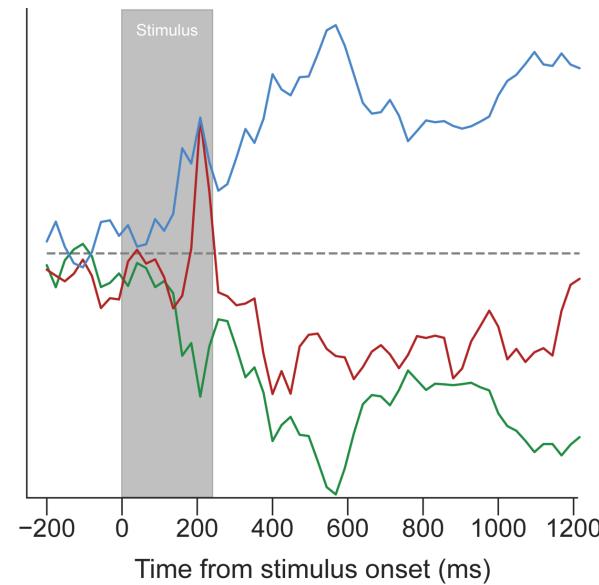


# Individual differences

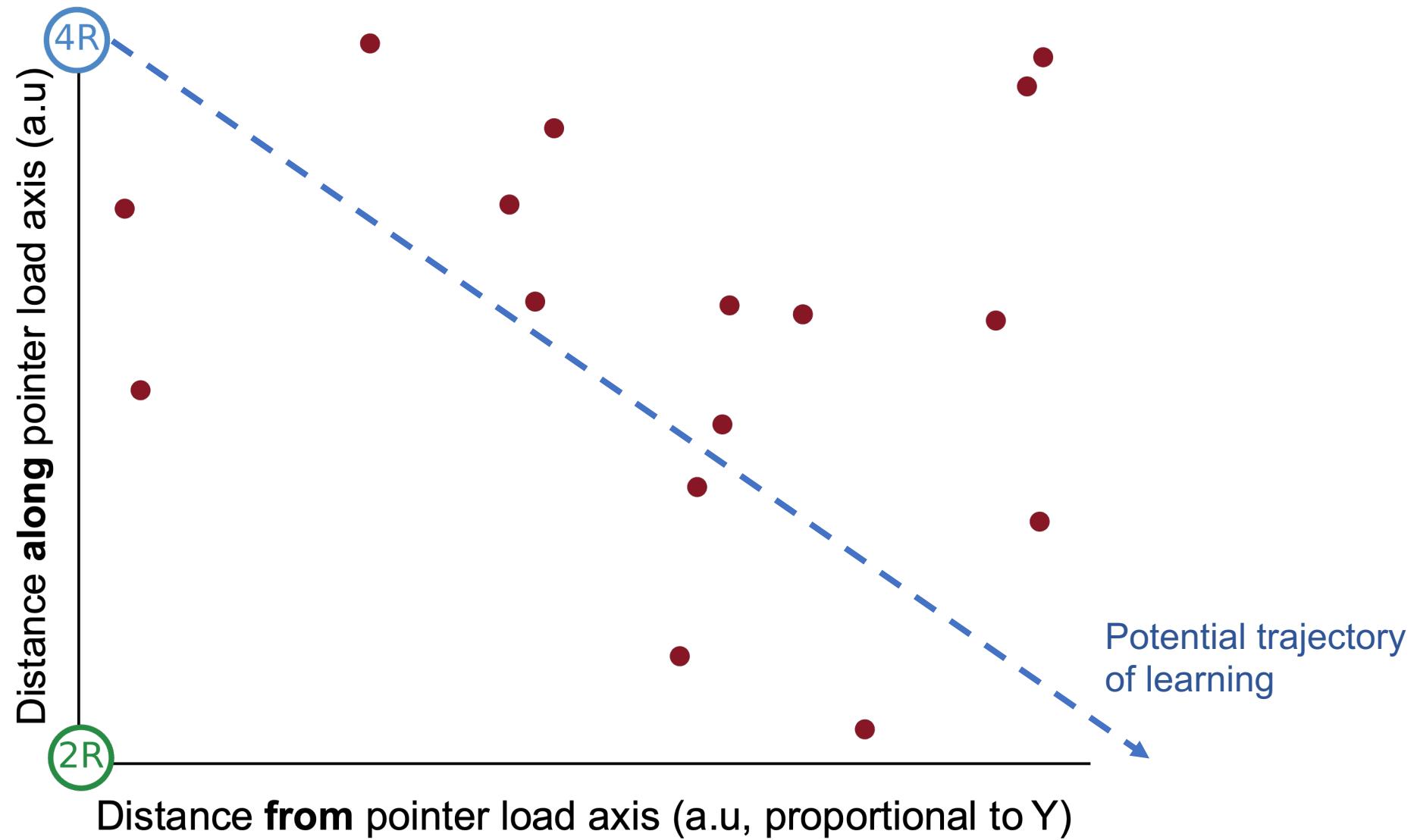
“Weak chunking”



“Strong chunking”



# Individual differences



# Experiment 2 - Training

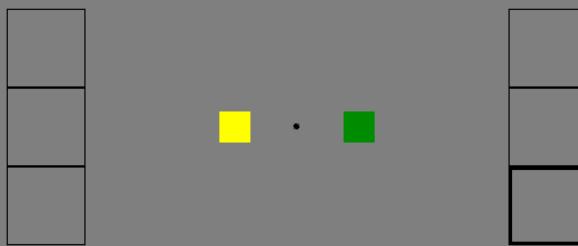
- Trained subjects to learn three color triplets



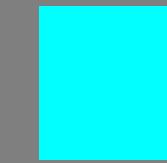
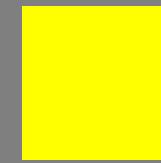
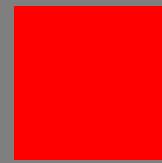
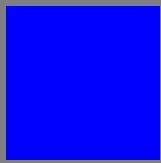
# Experiment 2 - Training



# Experiment 2 - Training

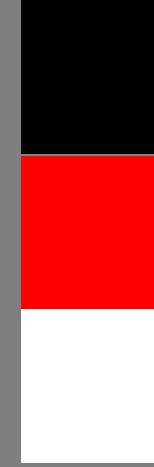


# Awareness Test



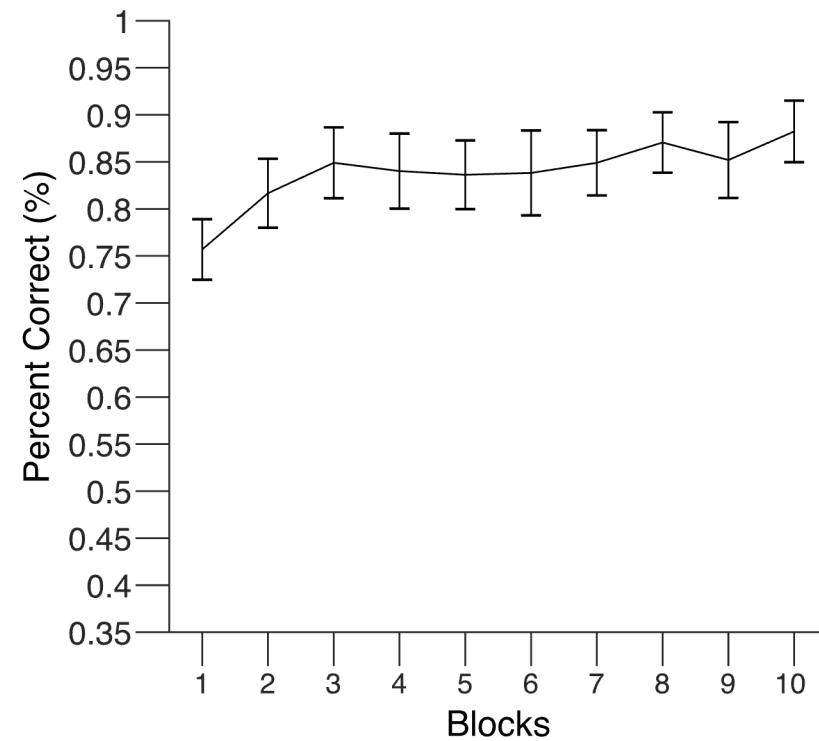
# Awareness Test

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

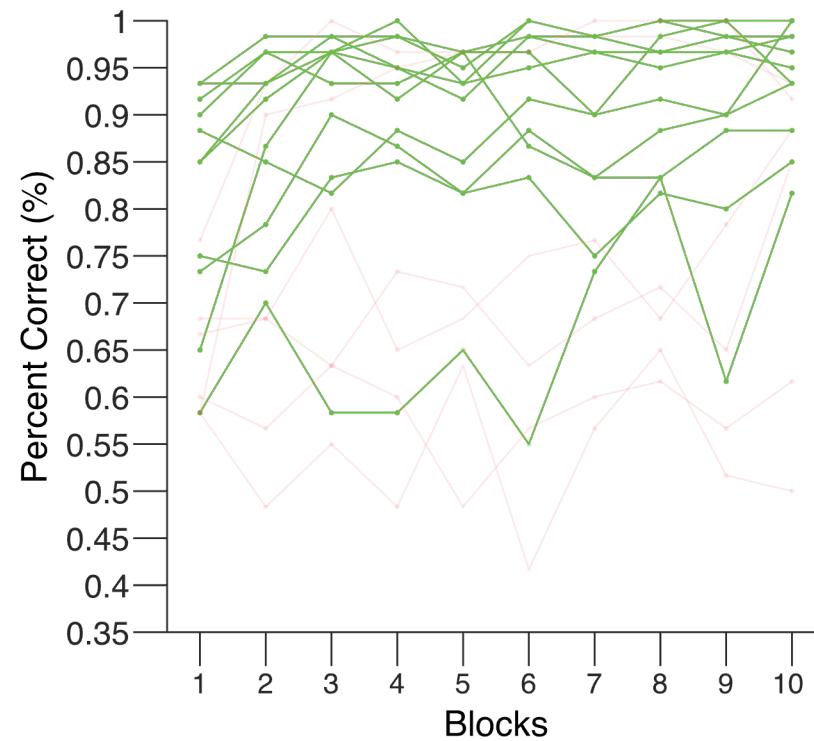


# Experiment 2 Training

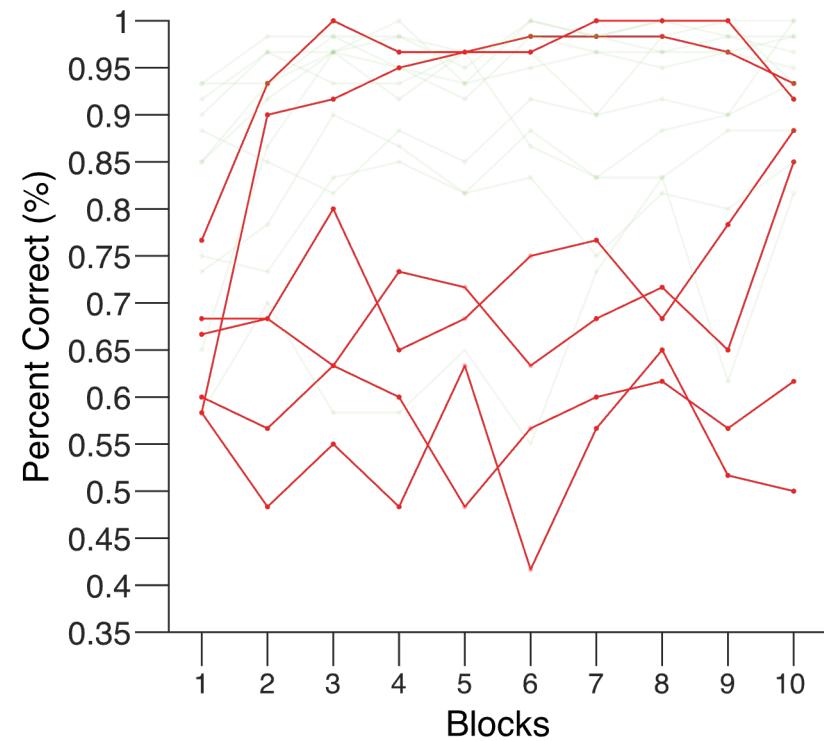
Aggregate



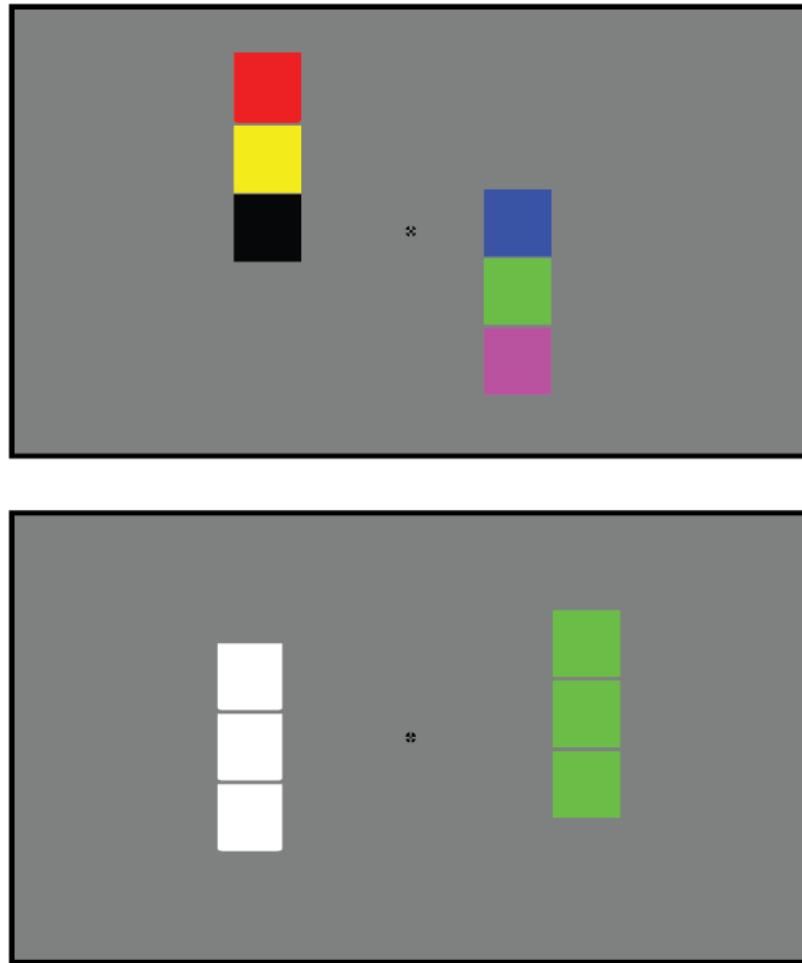
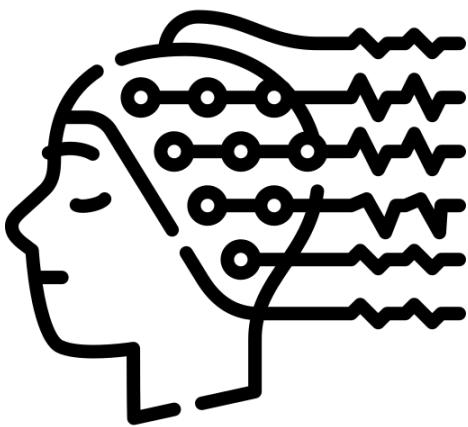
Learners



Non-learners



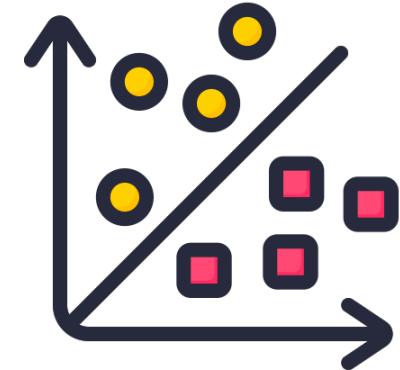
# Experiment 2: EEG session



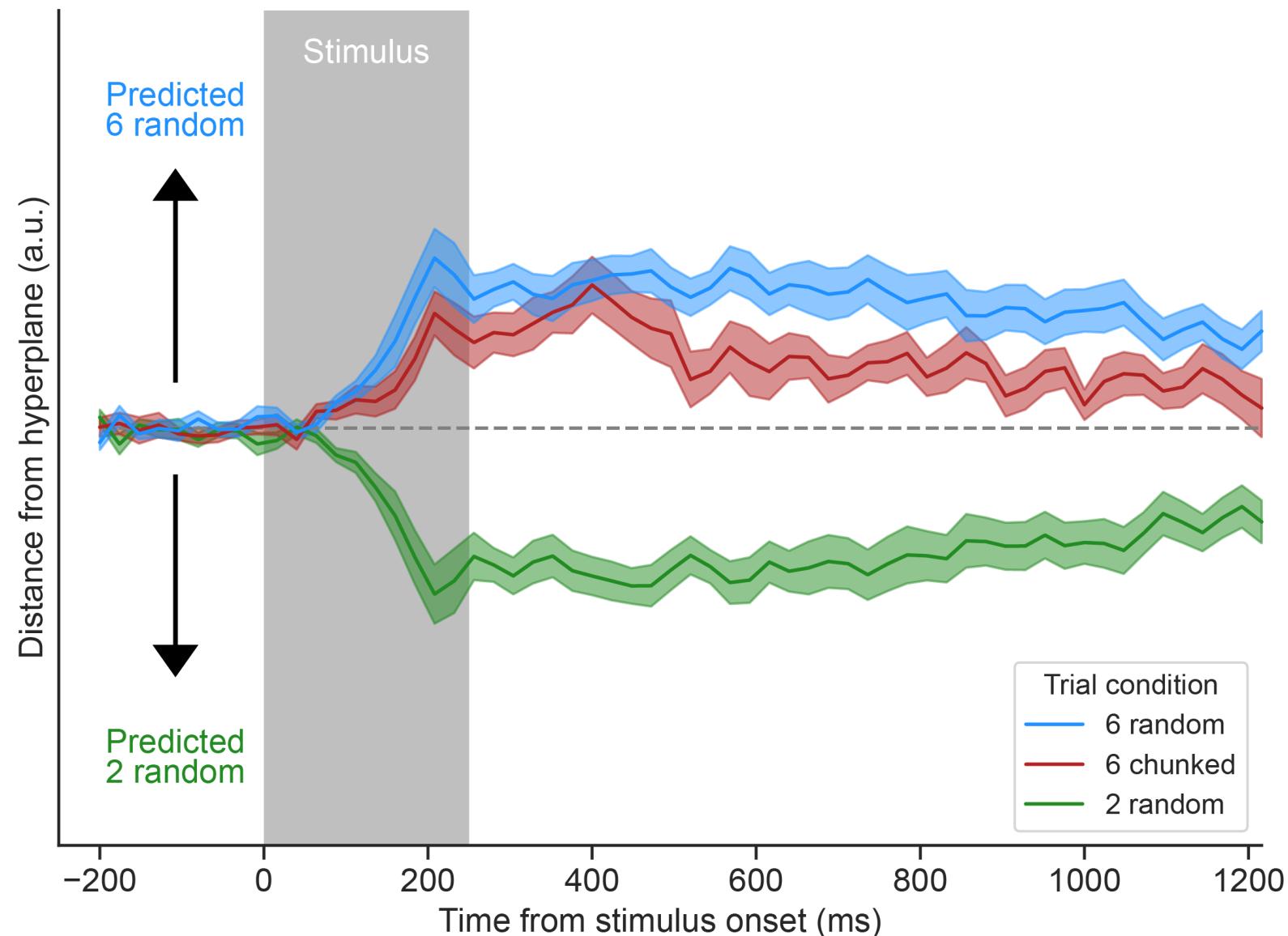
Six random  
Six chunked

Two random

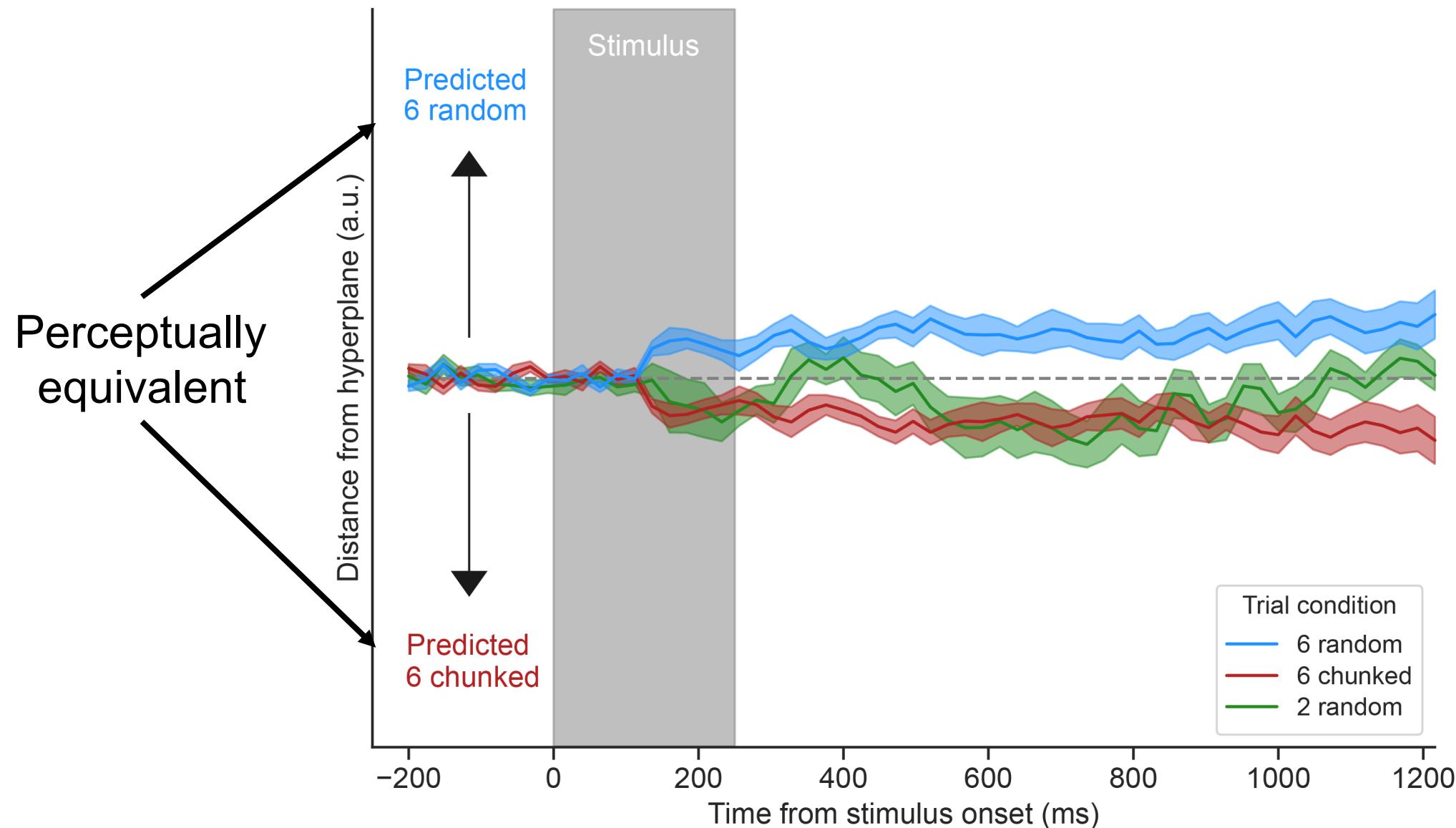
Perceptually equivalent



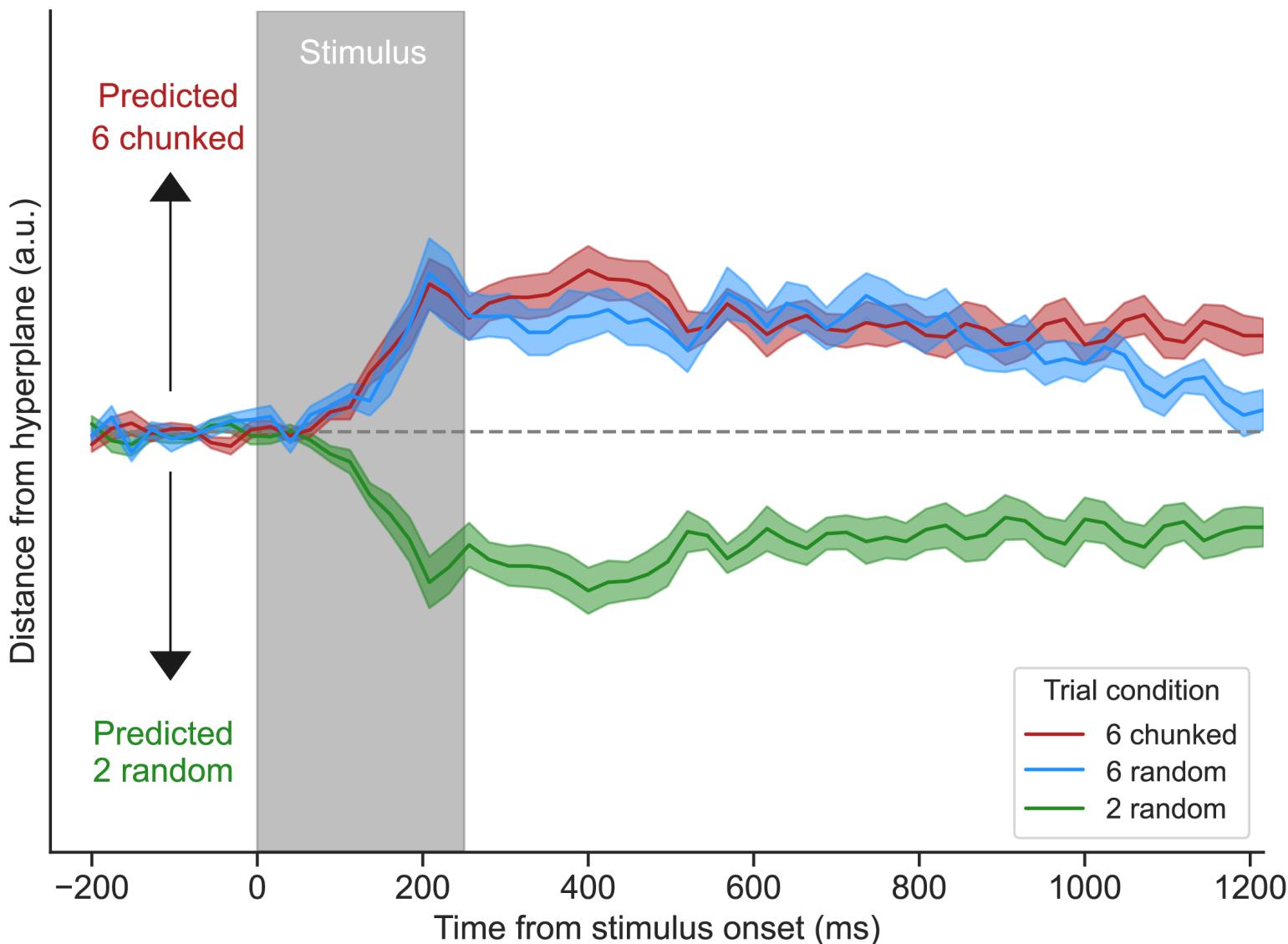
# Train 6 random versus 2 random, test 6 chunked



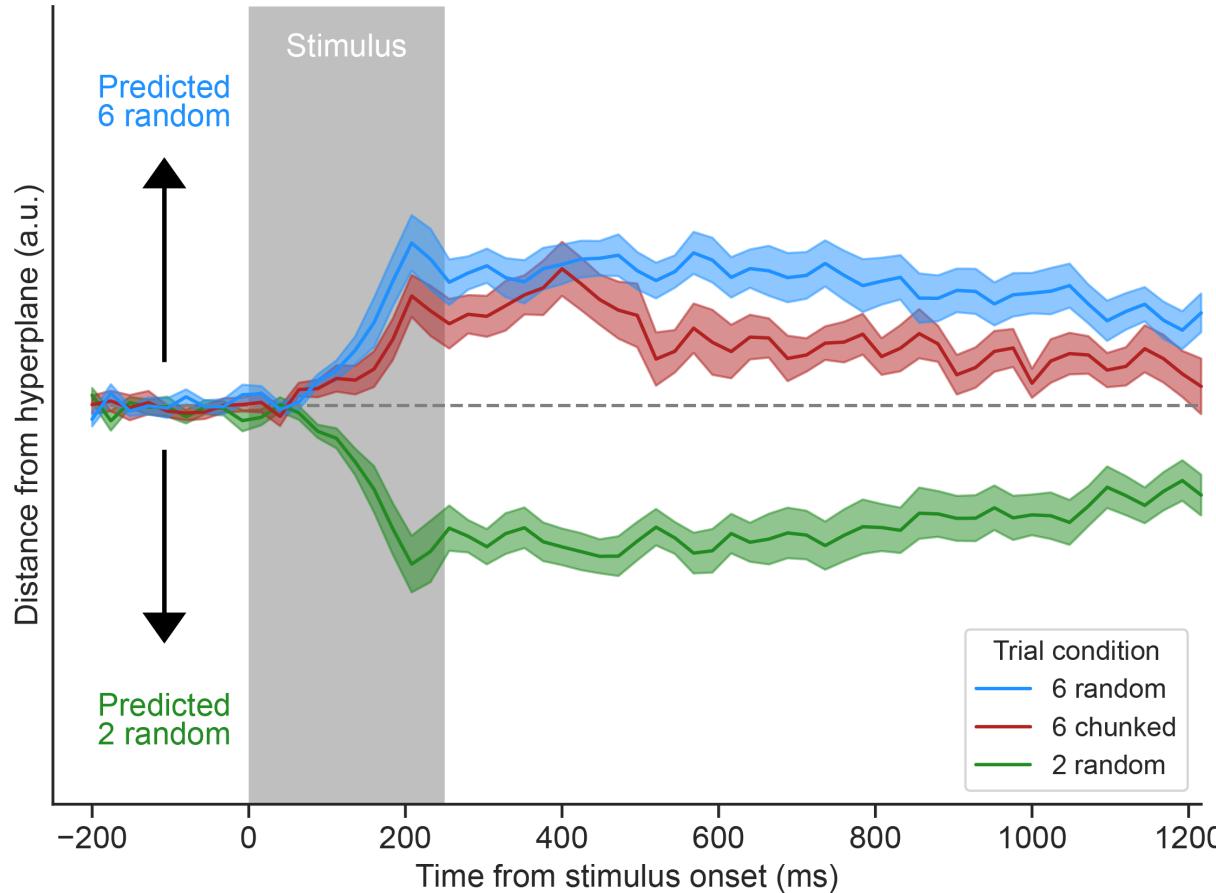
# Train 6 random versus 6 chunked, test 2 random



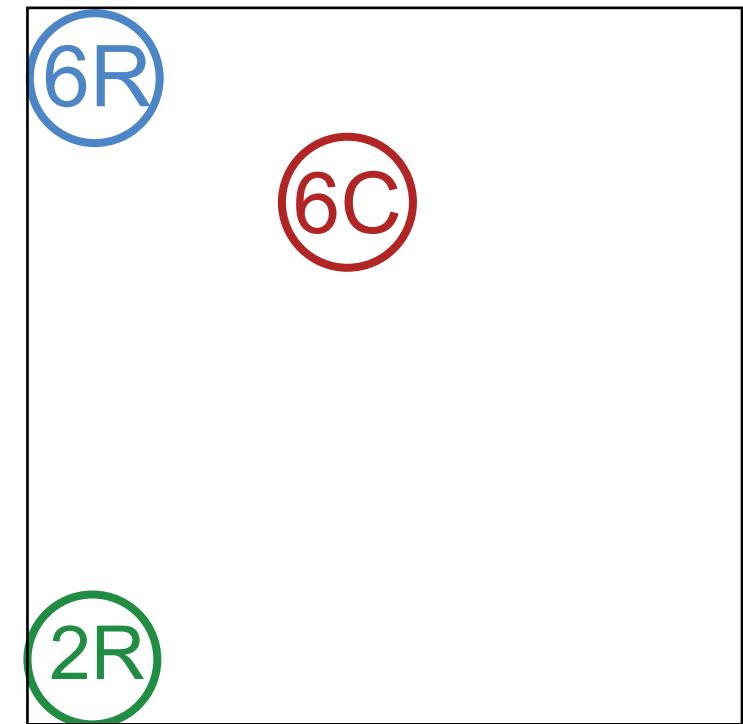
# Train 2 random versus 6 chunked, test 6 random



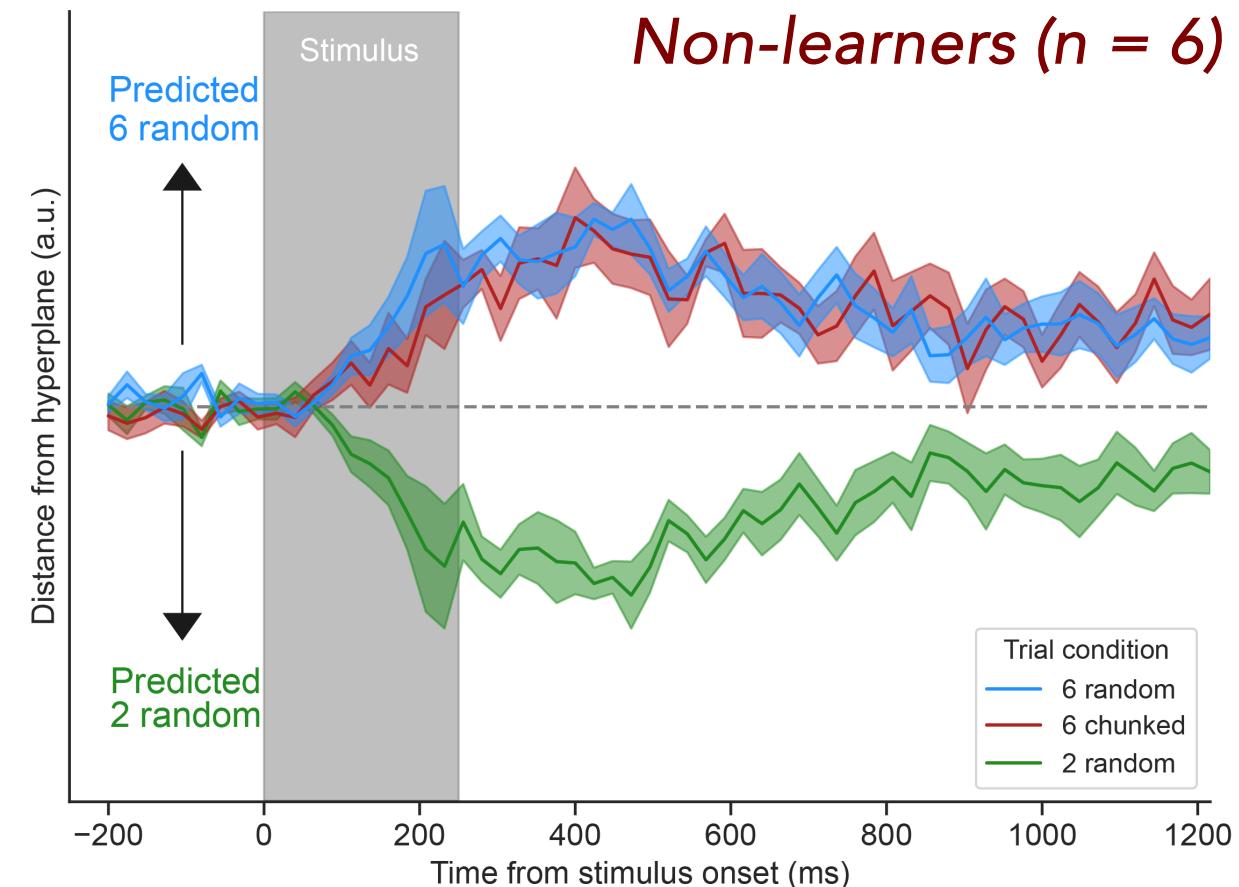
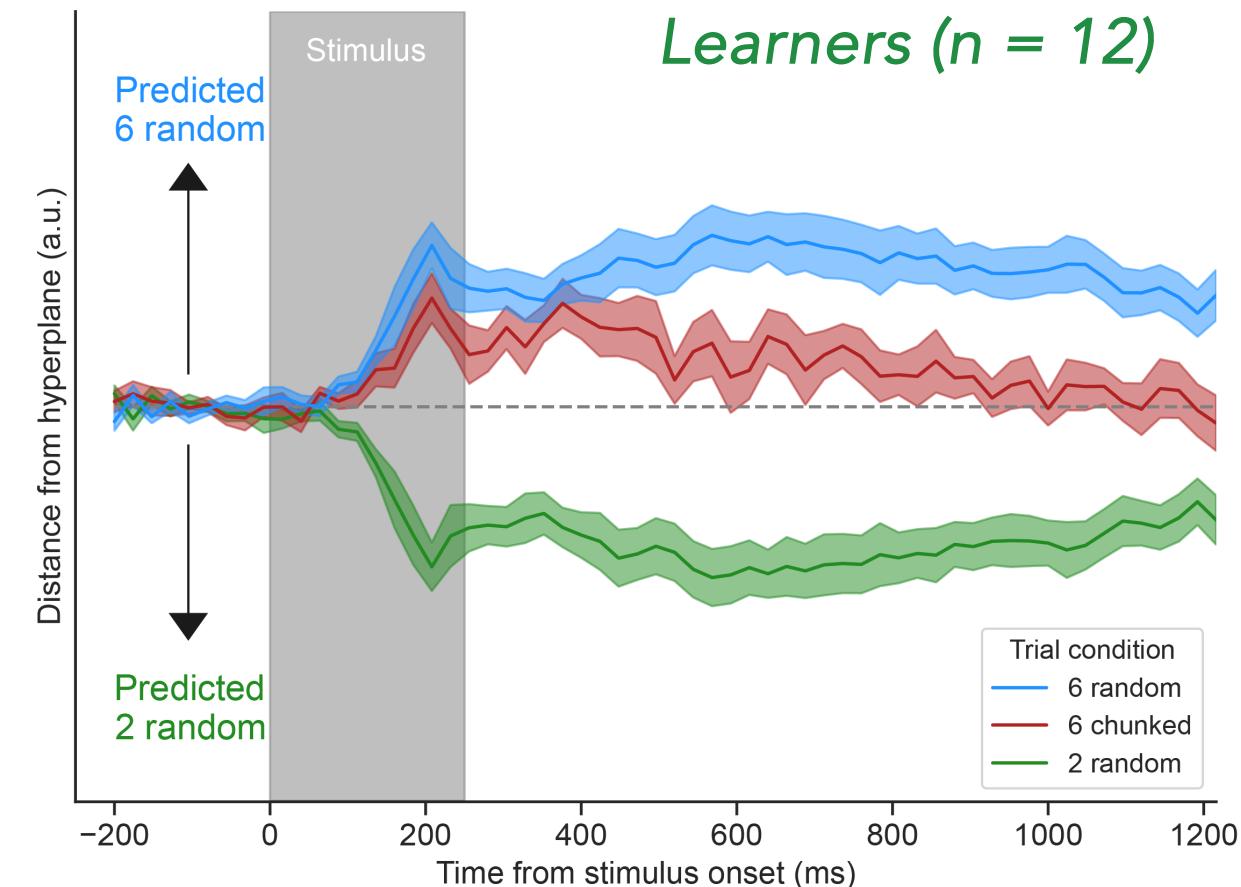
# Train 6 random versus 2 random, test 6 chunked



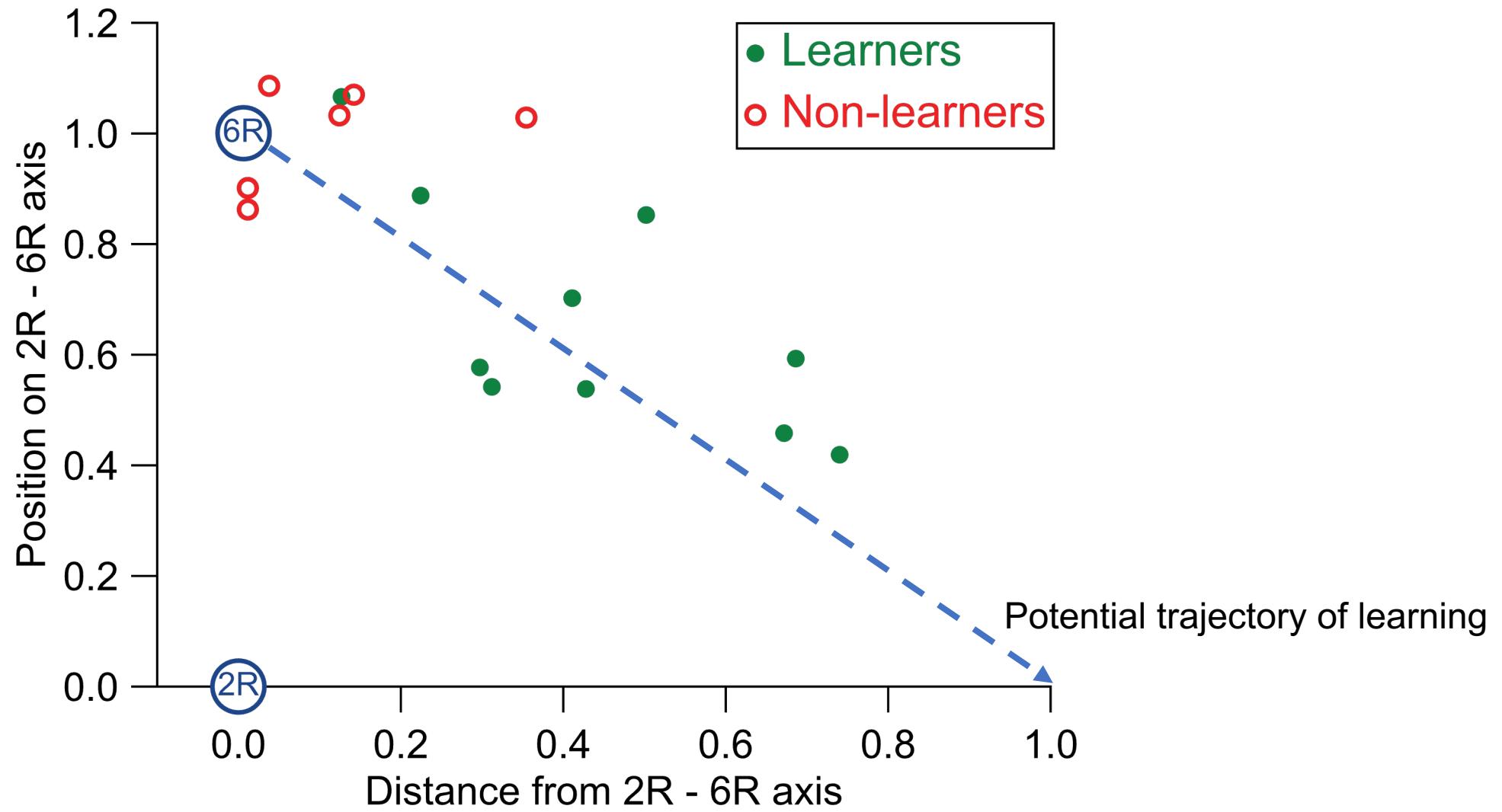
Multidimensional scaling



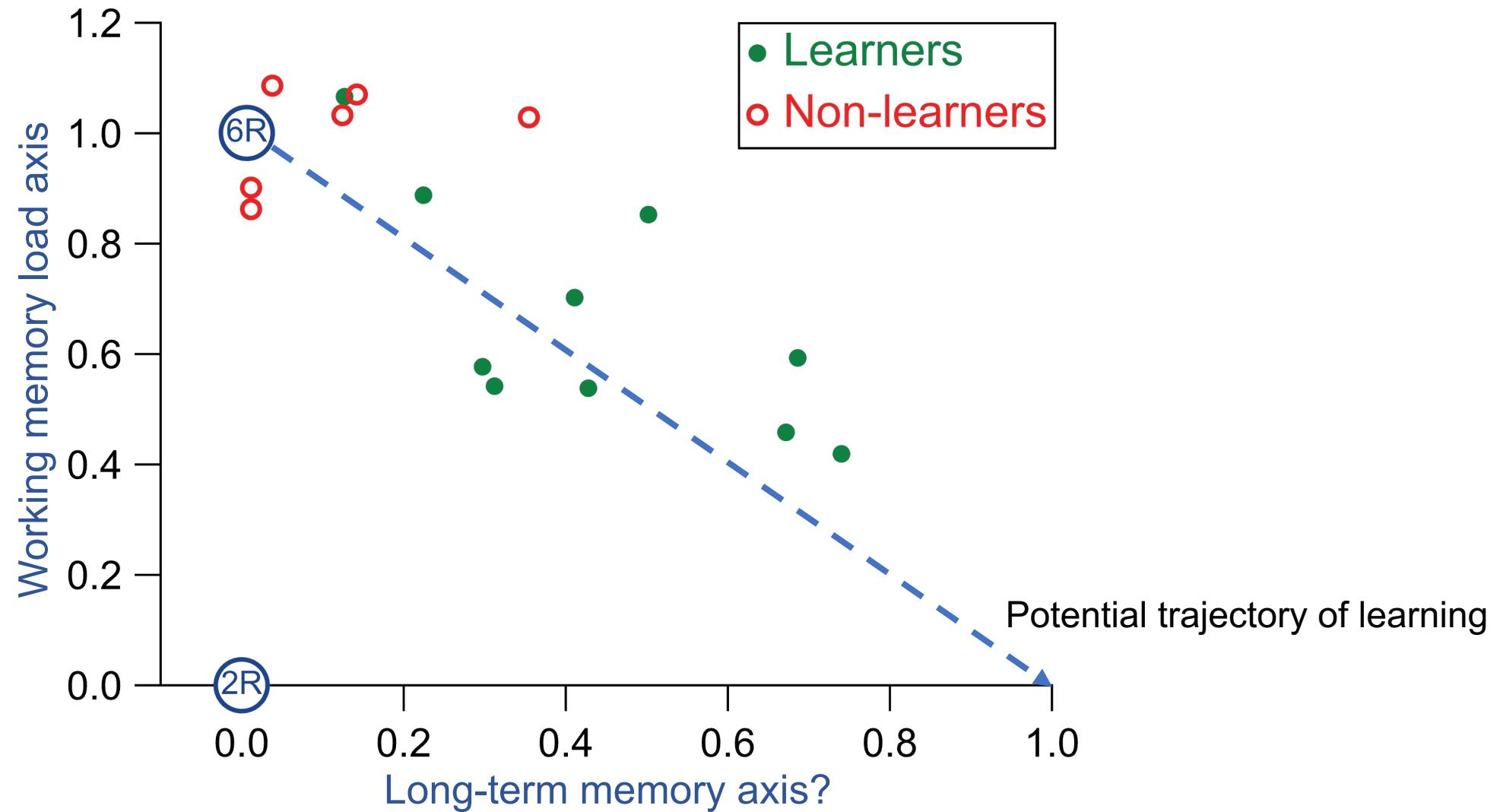
# Learners vs non-learners



# Multidimensional scaling on each subject



# Multidimensional scaling on each subject



# Conclusions

- We asked whether associative learning:
  - Increases the number of representations in working memory (memory compression)
  - Or reduces the number of items stored in working memory (chunking)
- 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 limits

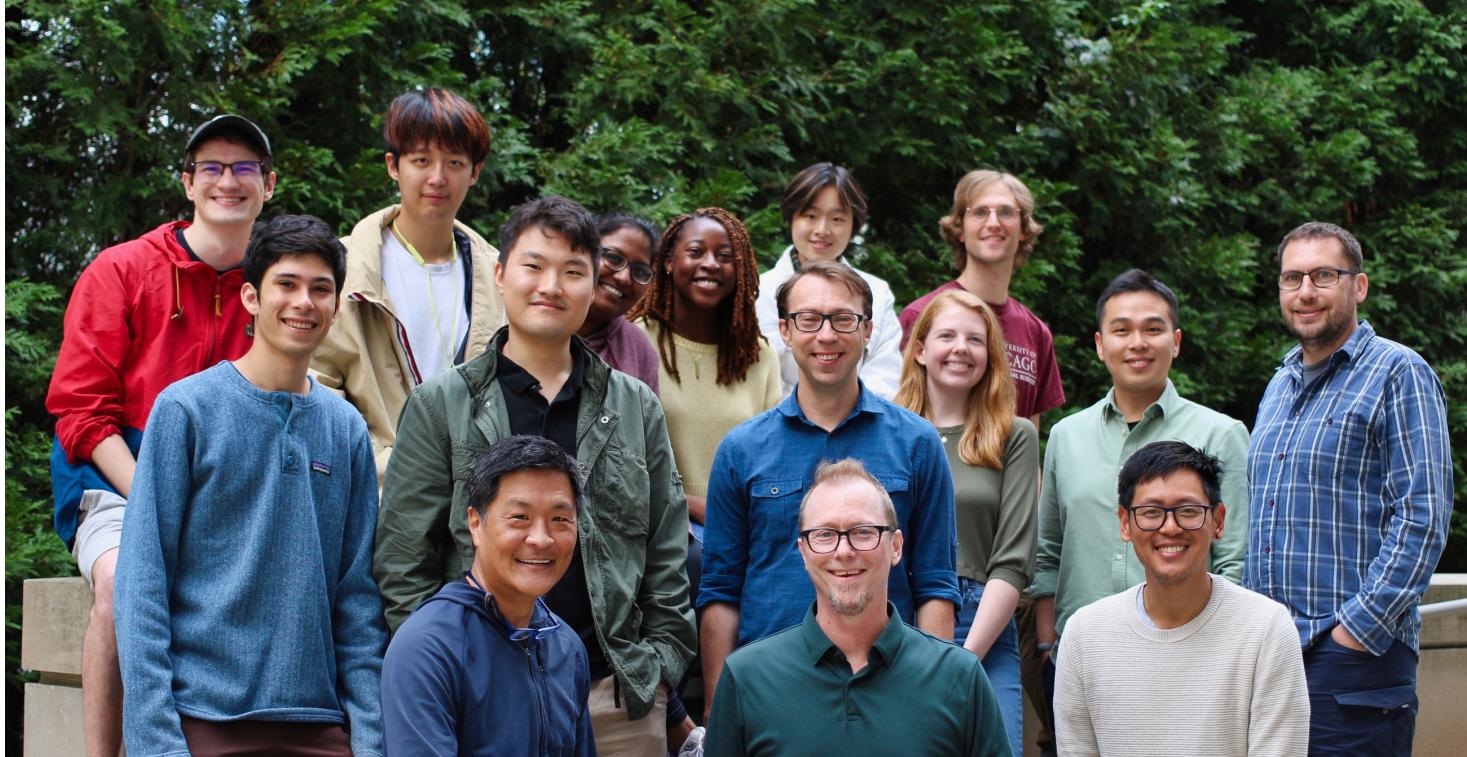


Will Epstein   Henry Jones   Darius Suplica



William Thyer   Edward Awh   Leo Chang

Check out Leo's related pre-data poster  
happening right now in Banyan



Scan for the Awh Vogel lab's VSS  
content including this talk's slides



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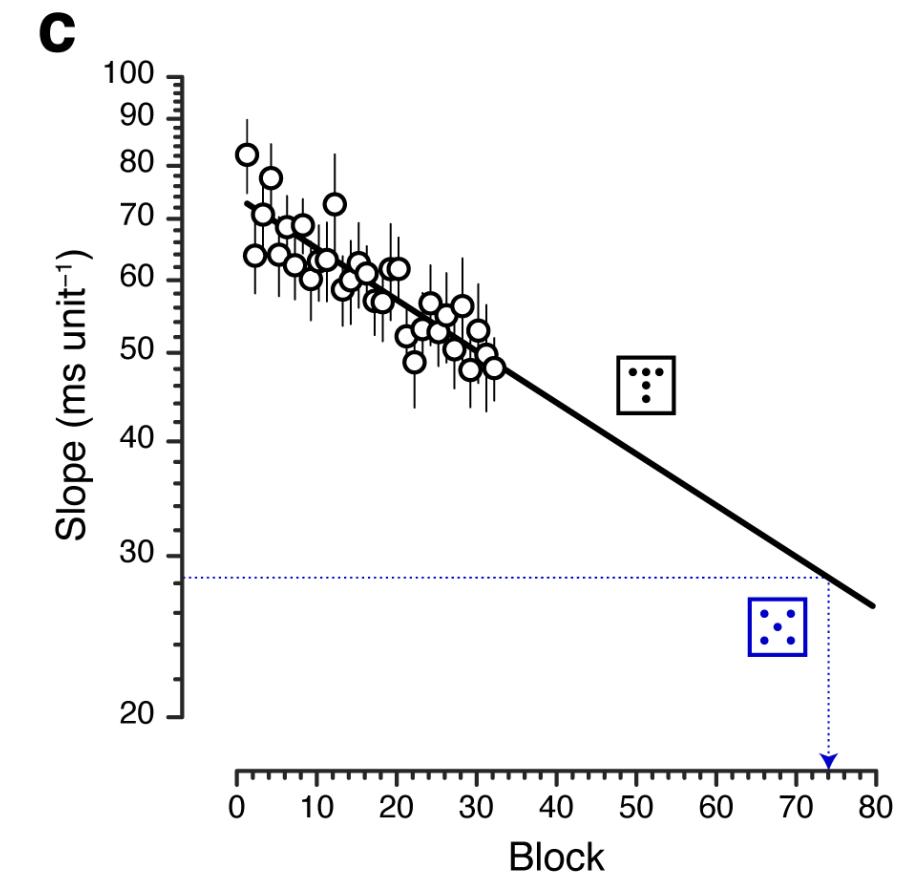
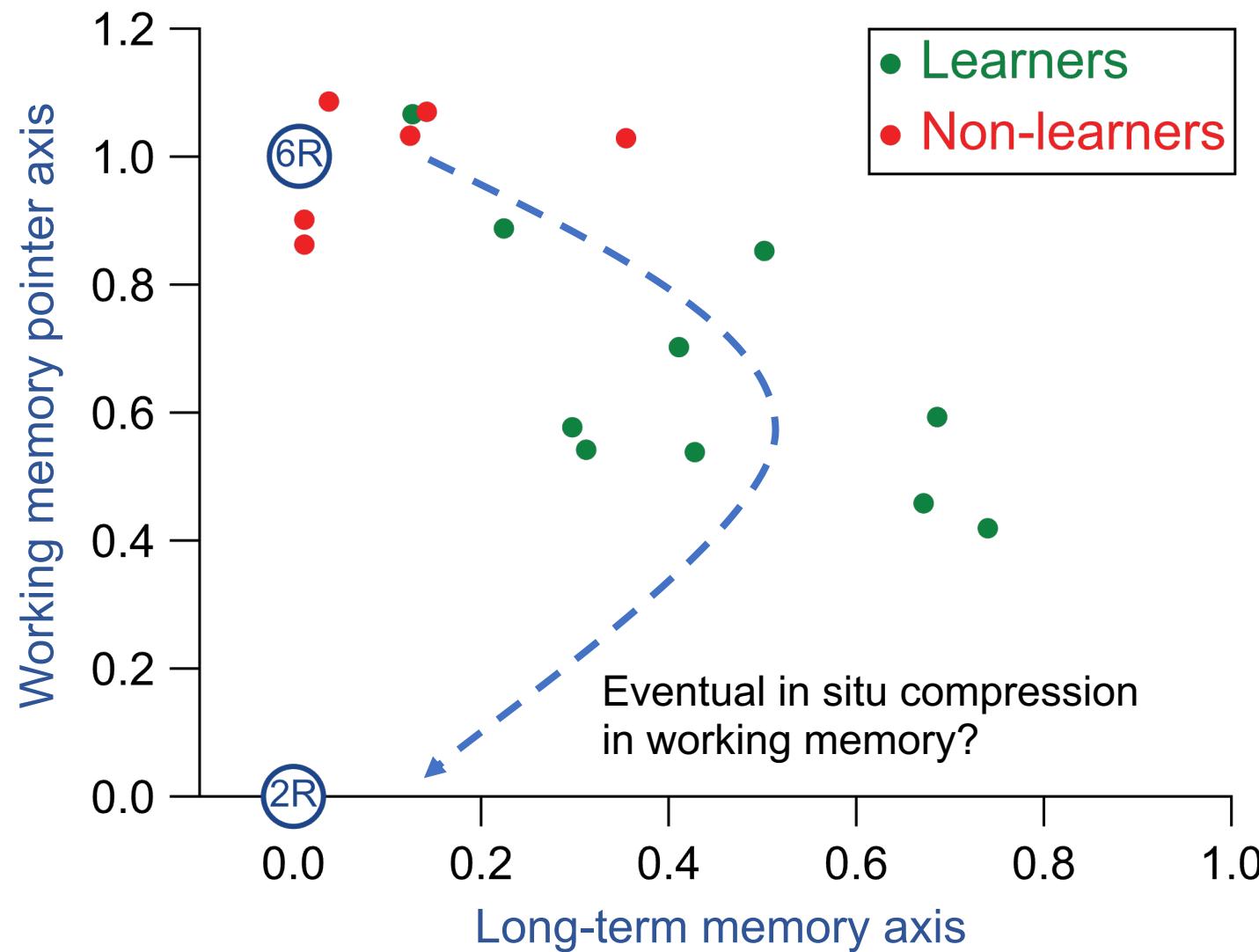


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# Why does the learned condition not cross the hyperplane?



# What are working memory pointers?

A ‘theory map’ of visual working memory (Ngiam, 2023)

