

Associative learning changes multivariate neural signatures of visual working memory

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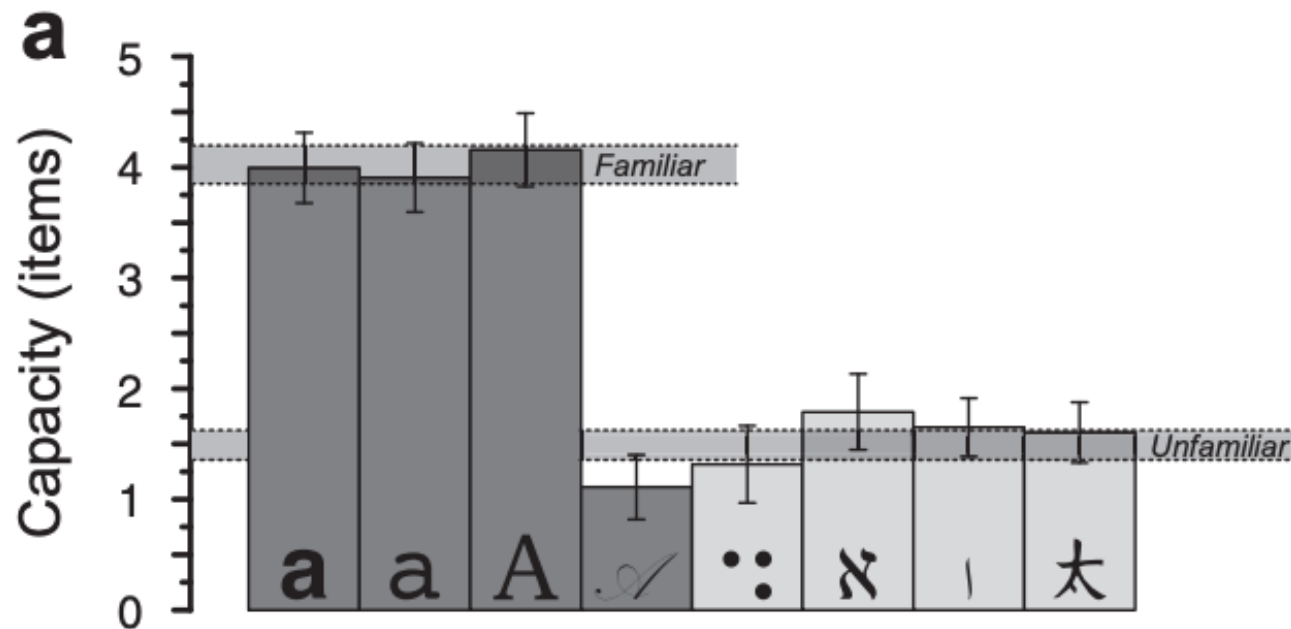


Edward Awh

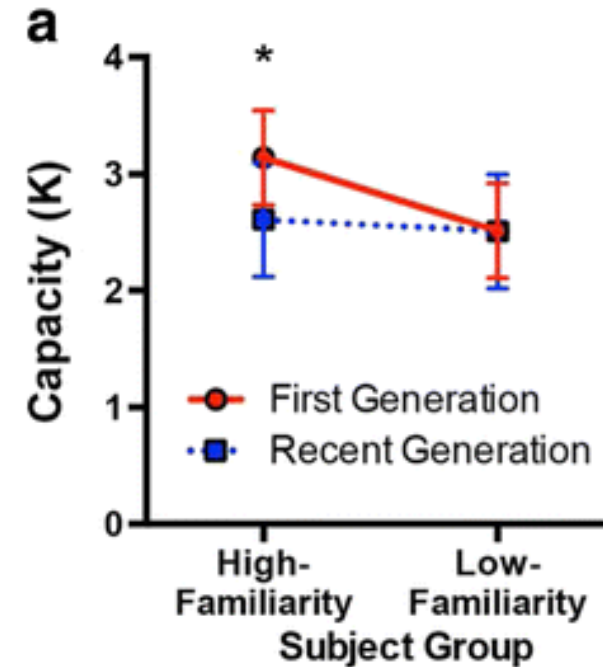


Working memory is aided by long-term memory

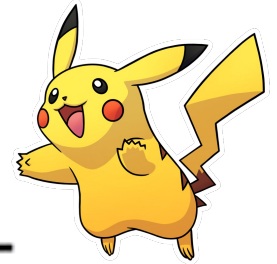
- A hallmark of our visual working memory system is its sharp capacity limit
- But this capacity limit can be overcome with **familiarity**:



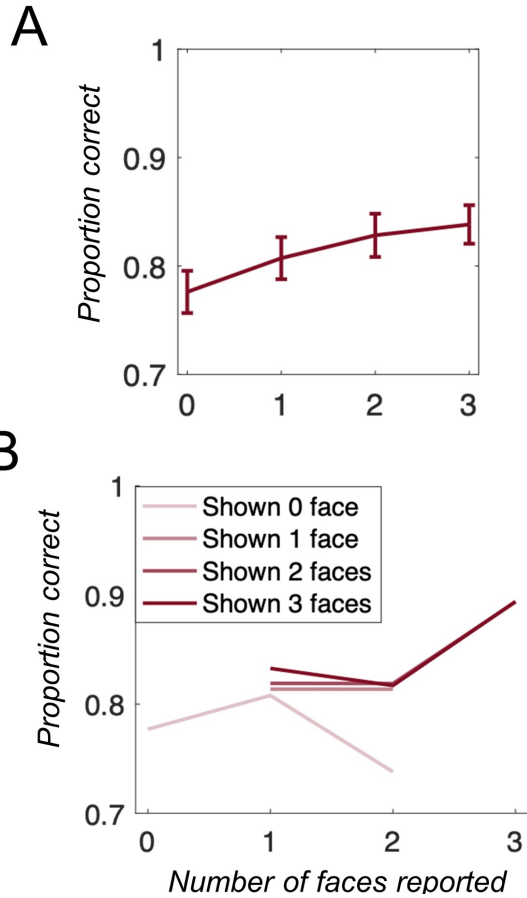
Ngiam et al. (2019) *JEP:G*



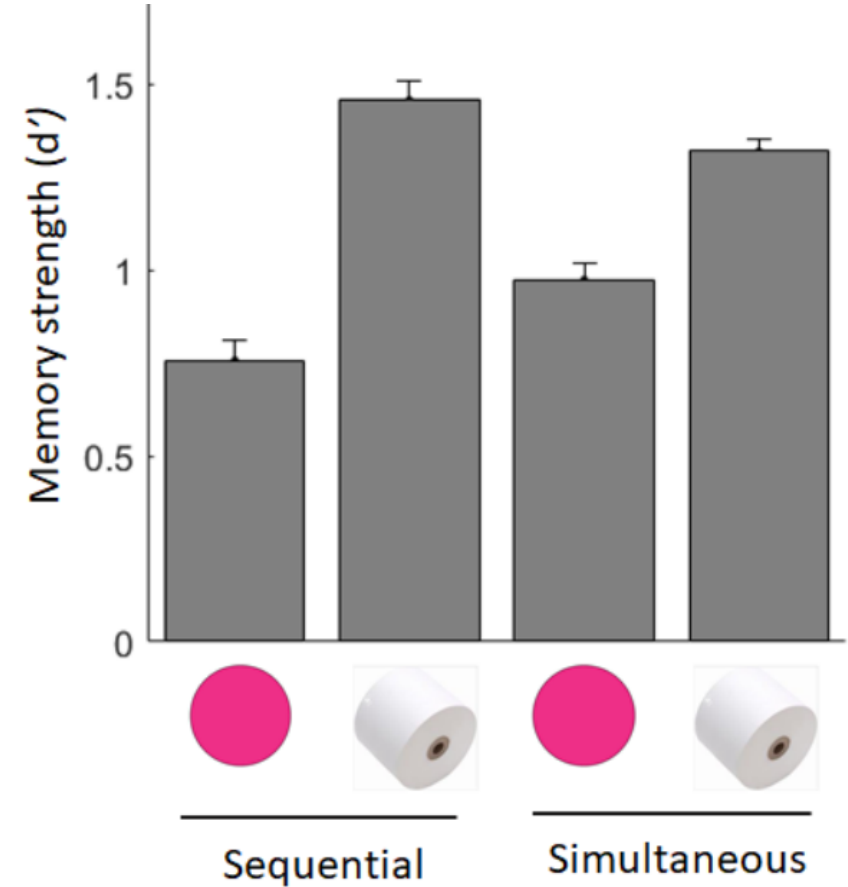
Xie and Zhang (2017) *M&C*



- Meaningfulness // real-world objects:

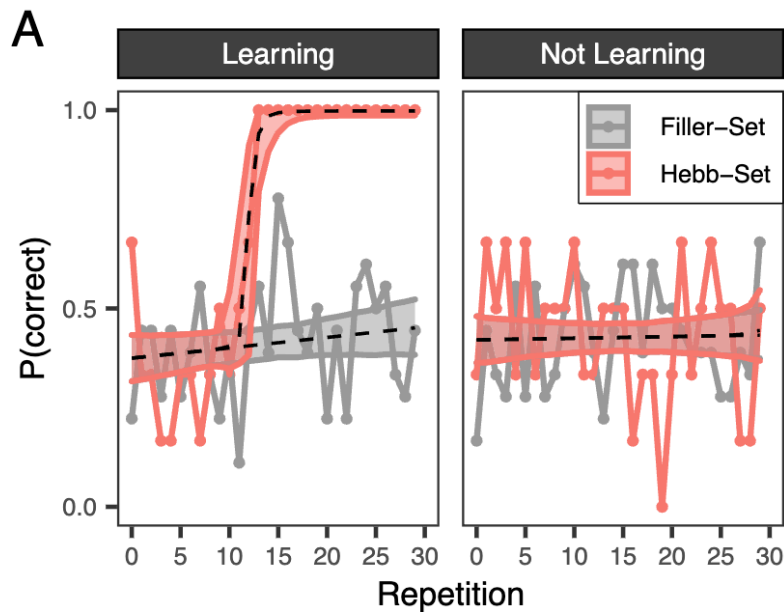
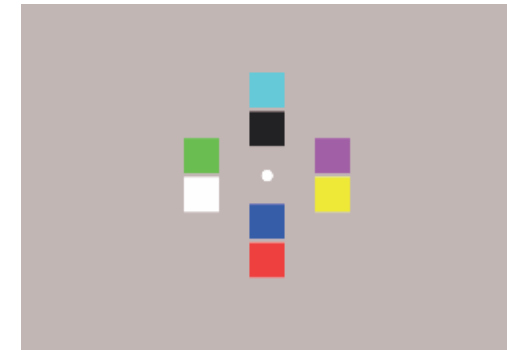
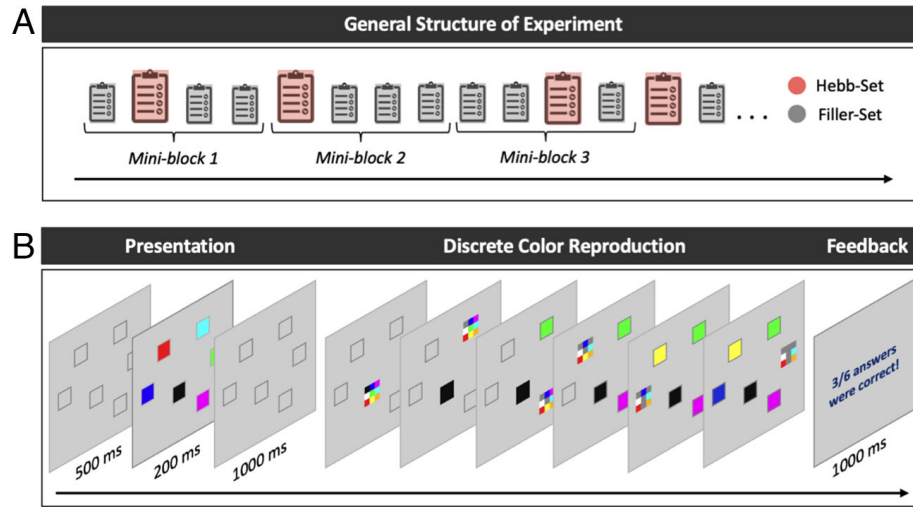


Asp et al. (2021) *JoCN*



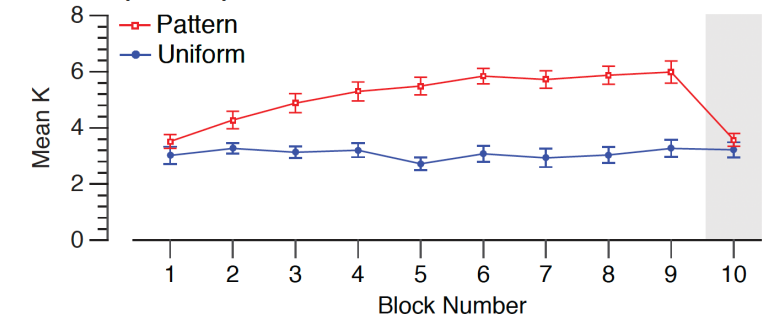
Brady and Störmer (2022) *JEP:LMC*

• Repetition learning // associative learning:

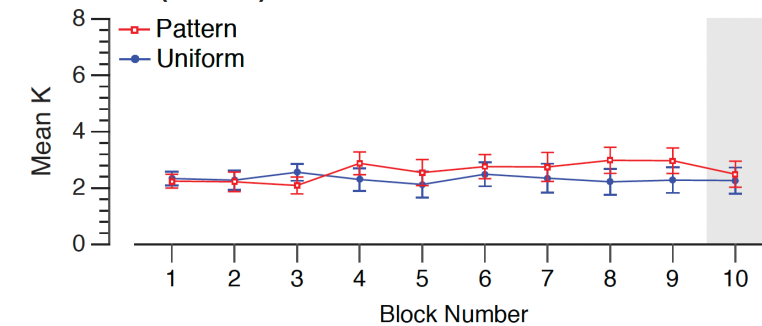


Musfeld et al. (2023) *PNAS*

Aware (n = 19)



Unaware (n = 13)



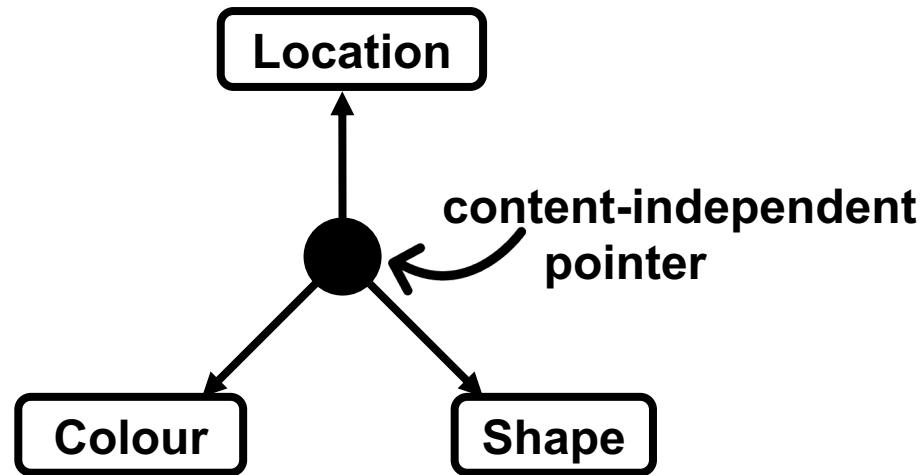
Ngiam et al. (2019) *JEP:G*

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

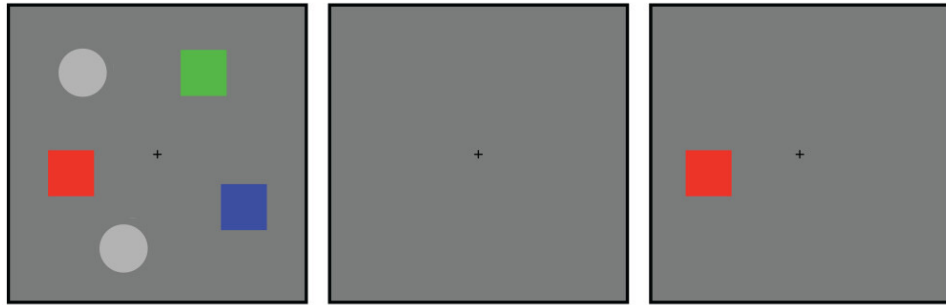
Pointers in working memory

- 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*

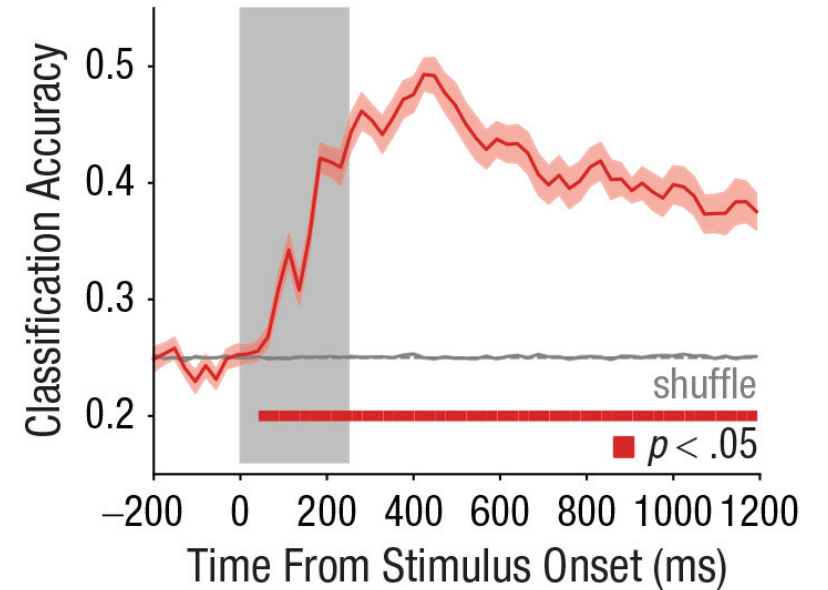


Multivariate neural signature of WM pointers

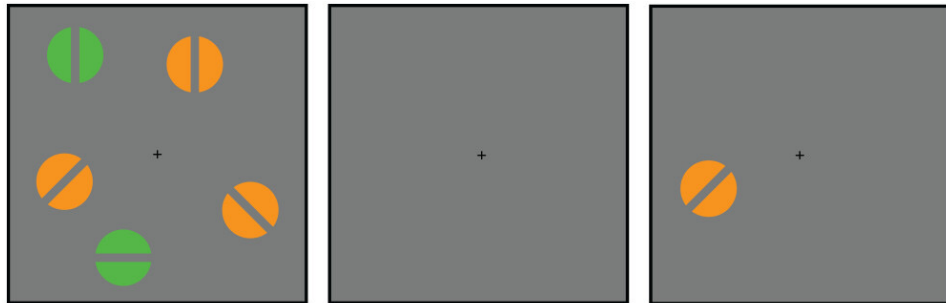
Experiment 1: Color



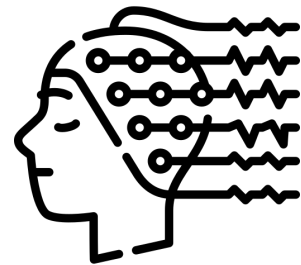
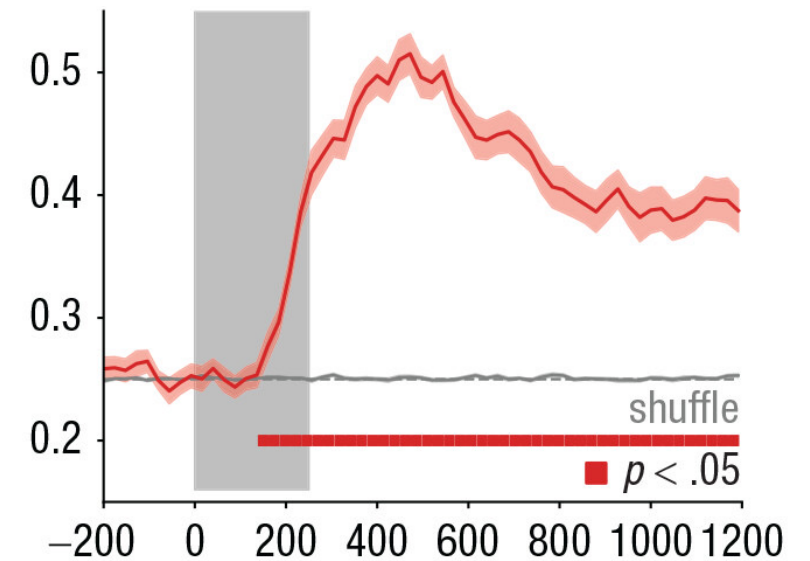
Train and test



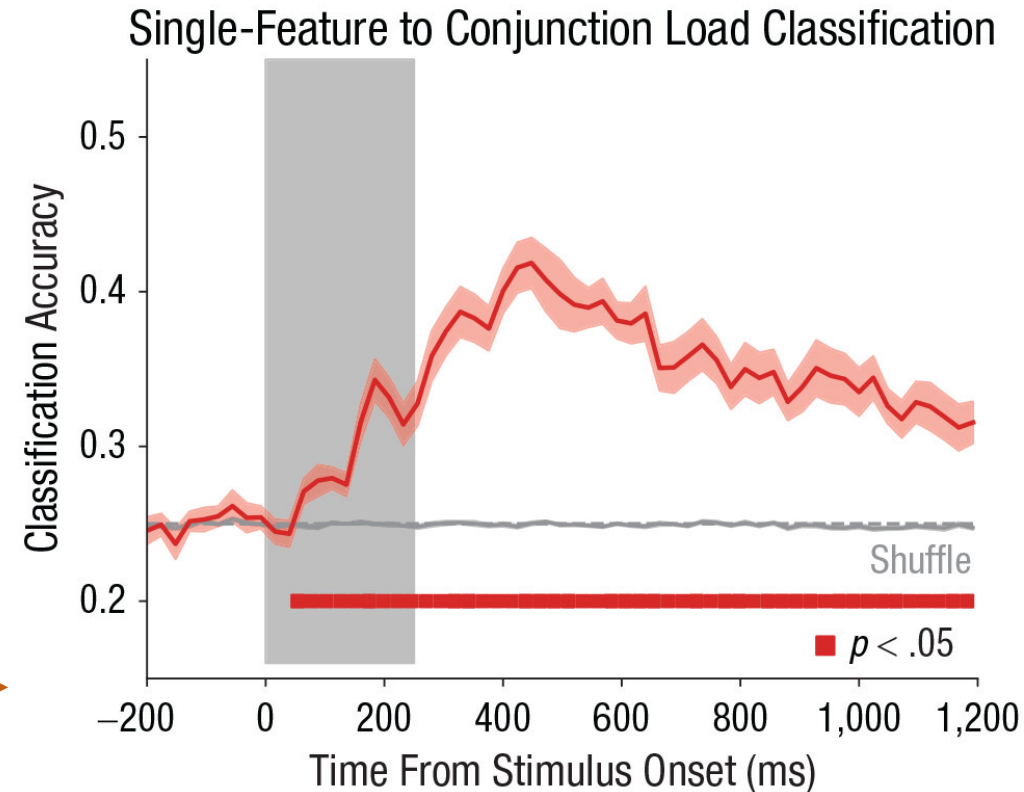
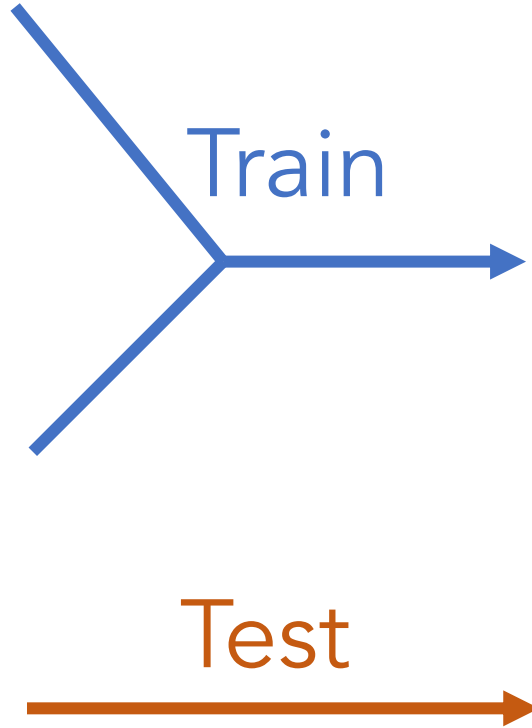
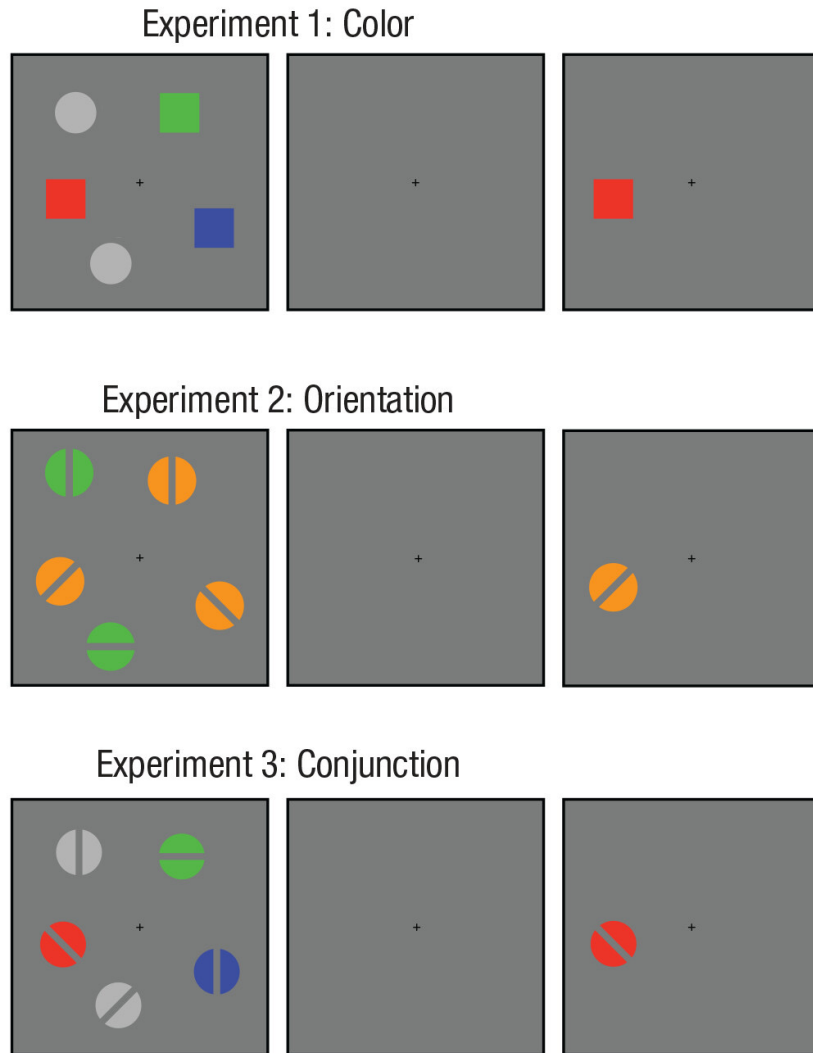
Experiment 2: Orientation



Train and test

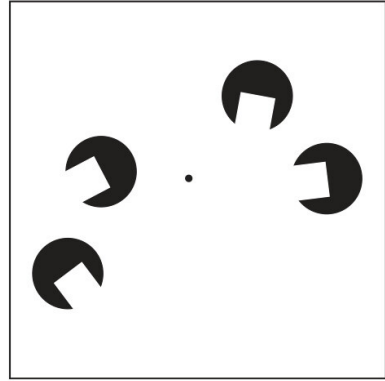


Multivariate neural signature of WM pointers

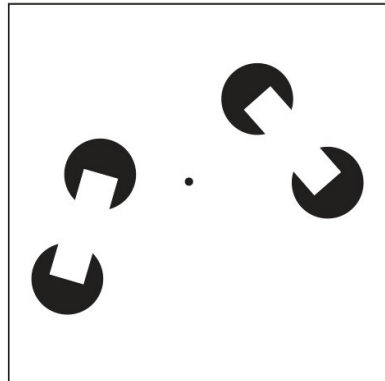


Multivariate neural signature of WM pointers

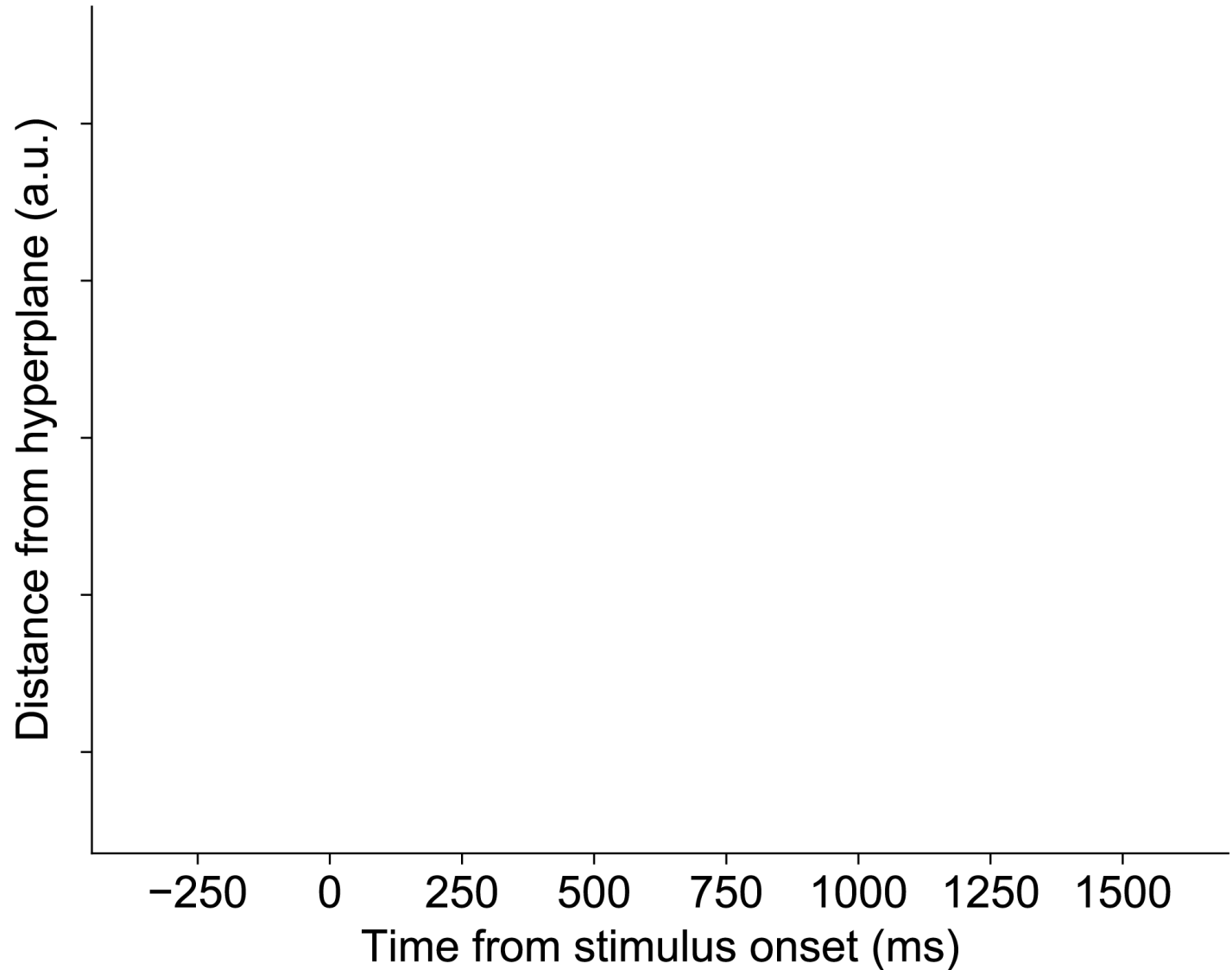
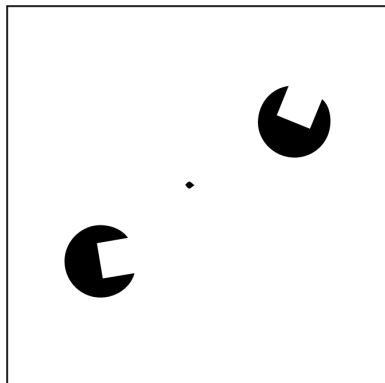
4 Ungrouped



4 Grouped

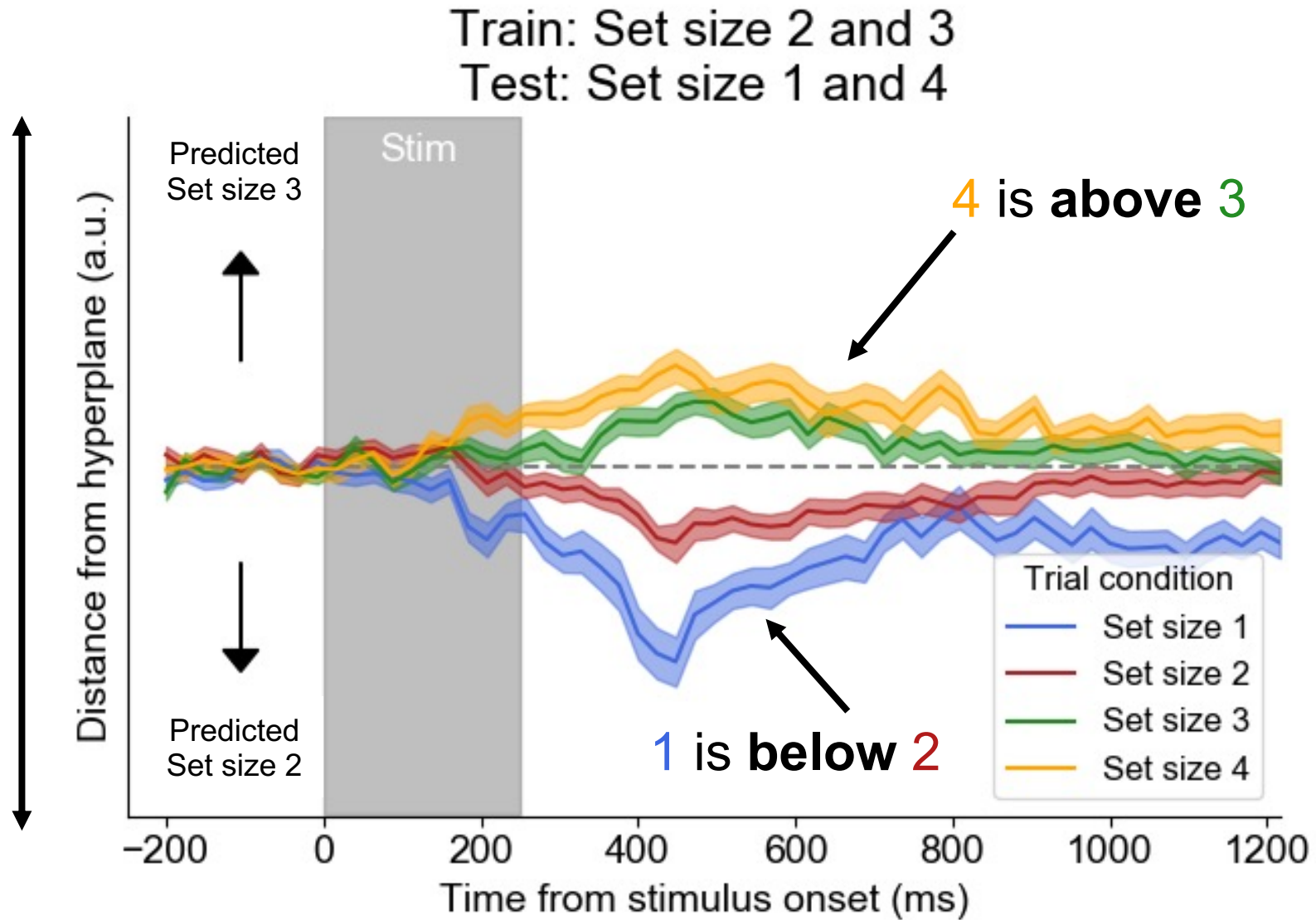


2 Ungrouped



Different working memory loads on the hyperplane

Working memory load signal



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

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

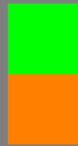


Darius Suplica

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)



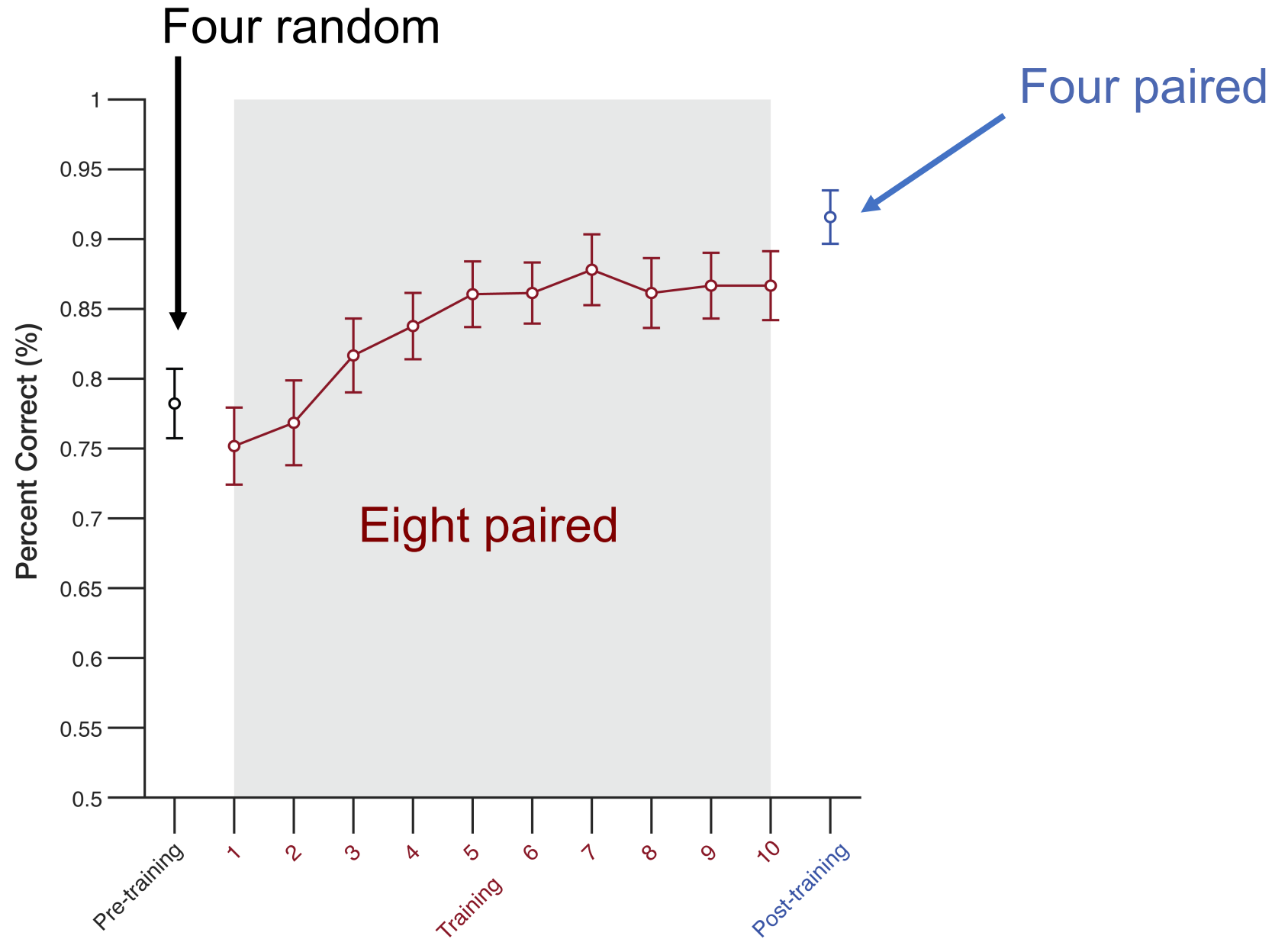
Experiment 1: Pre-training and post-training



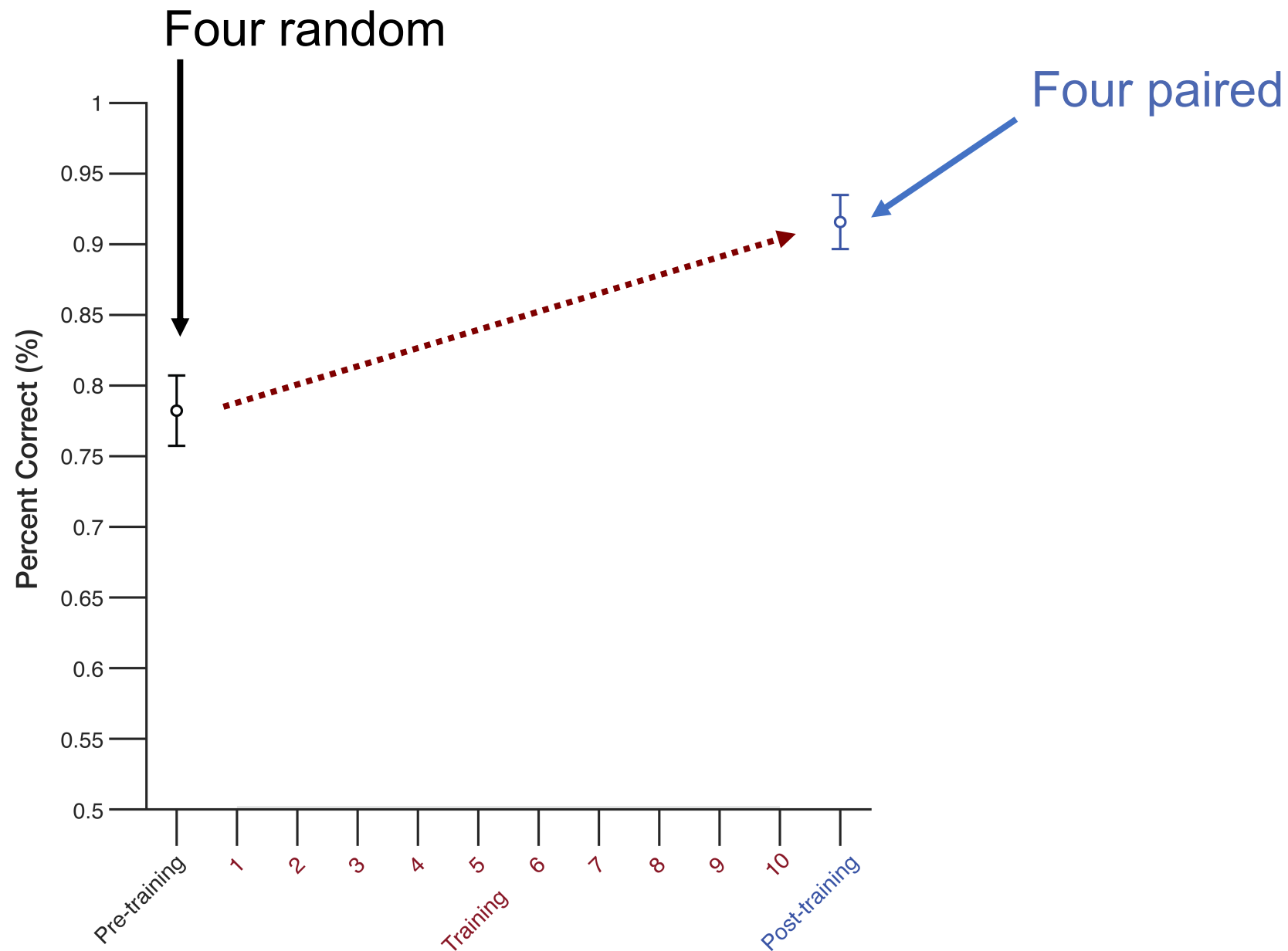
Experiment 1: Pre-training and post-training



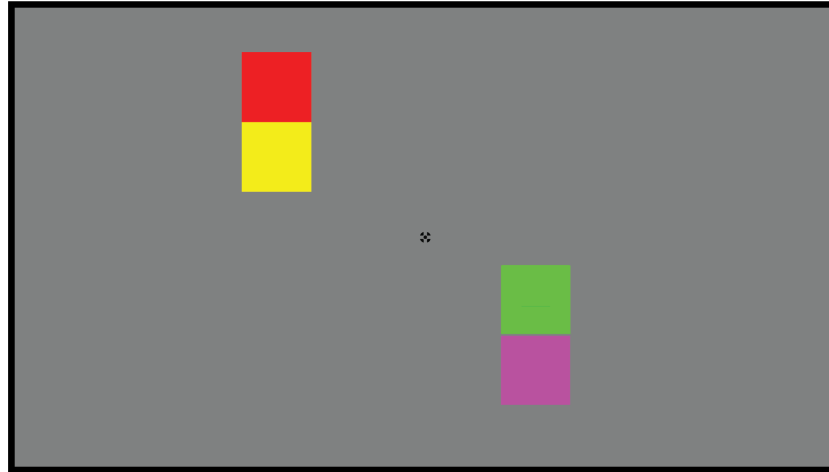
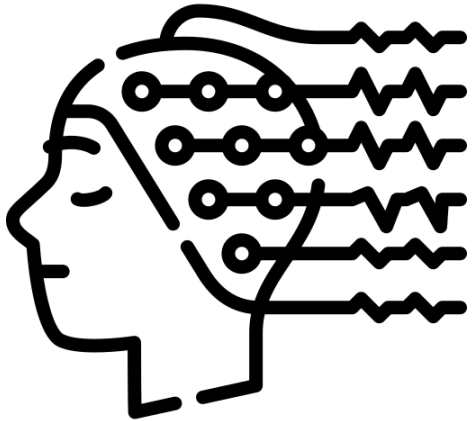
E1: training session – aggregate performance



E1: training session – average performance

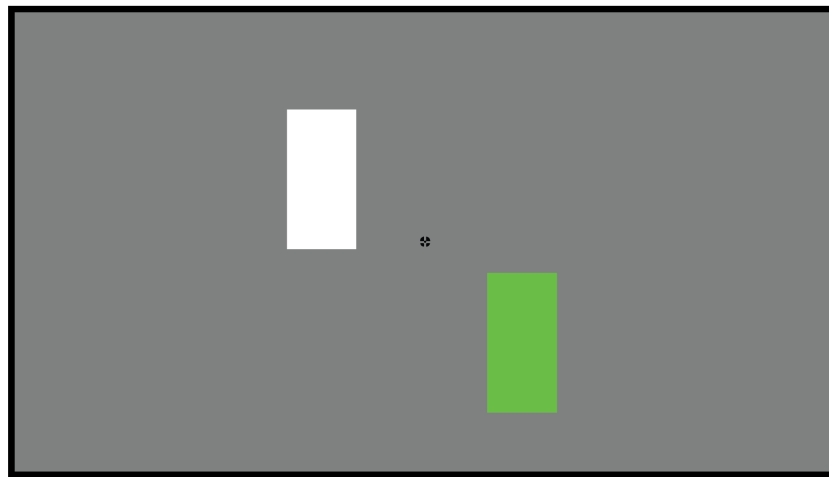
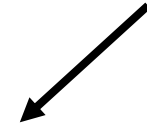


Experiment 1: EEG session



Four random
Four paired

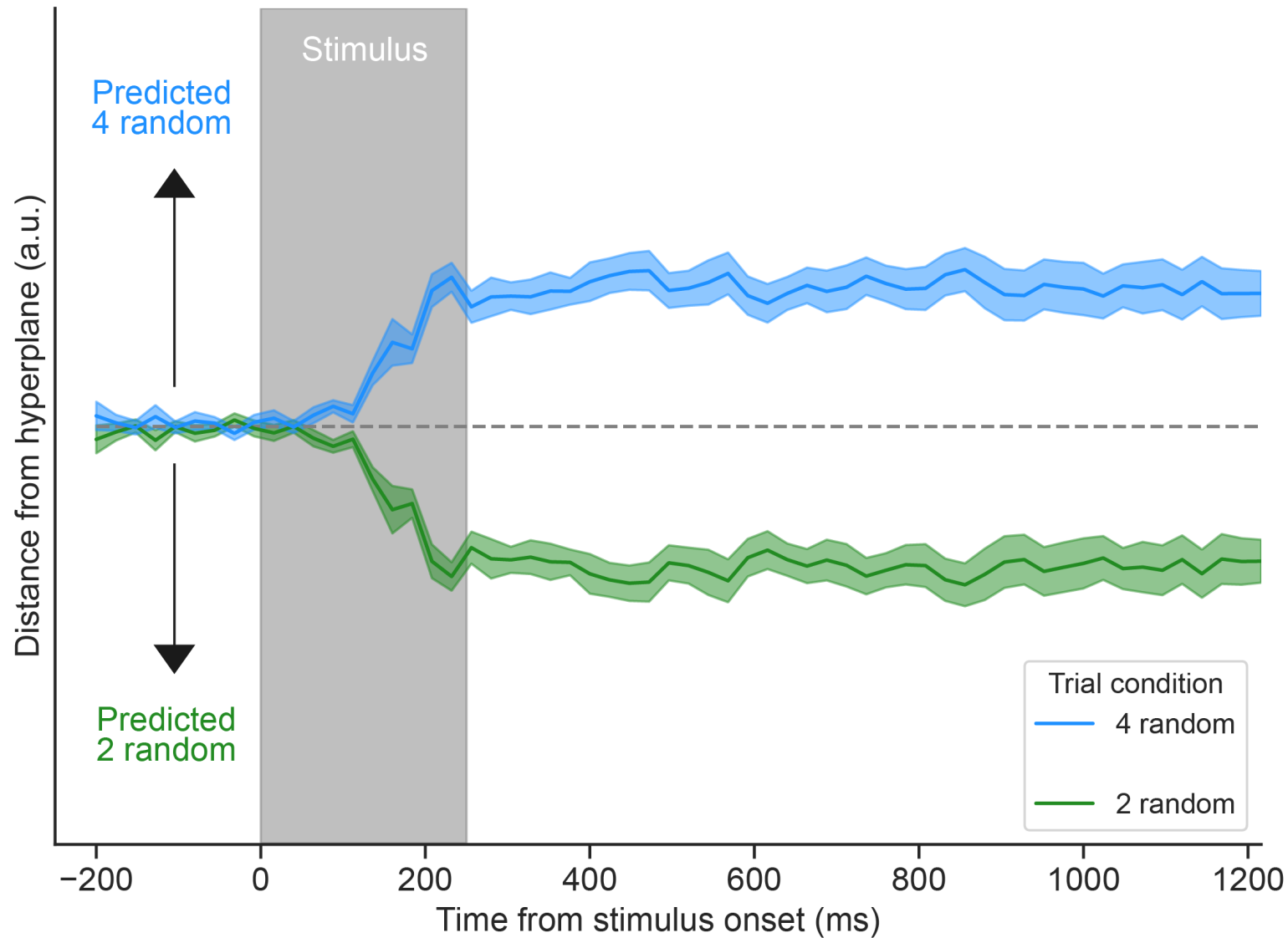
Perceptually
equivalent



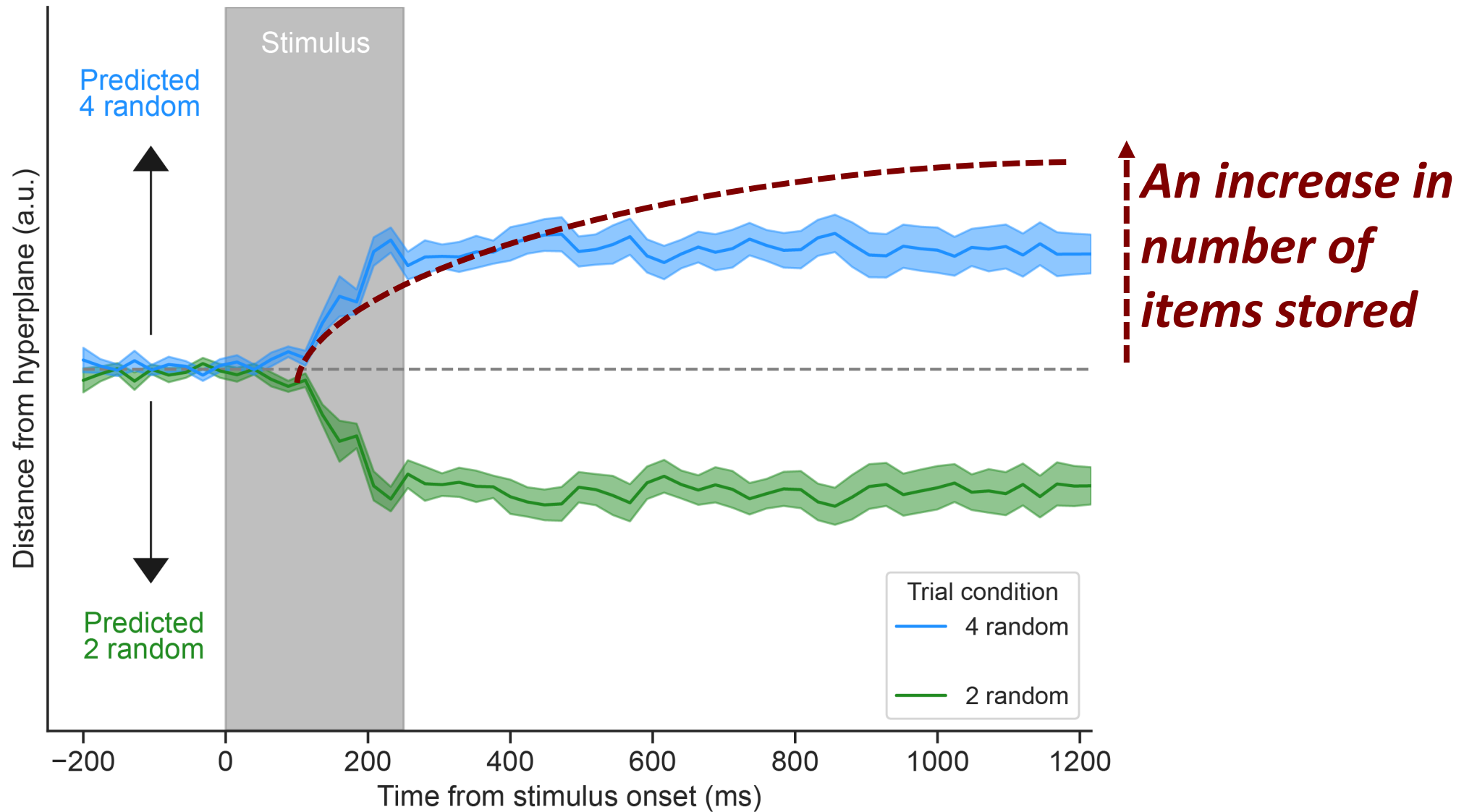
Two random



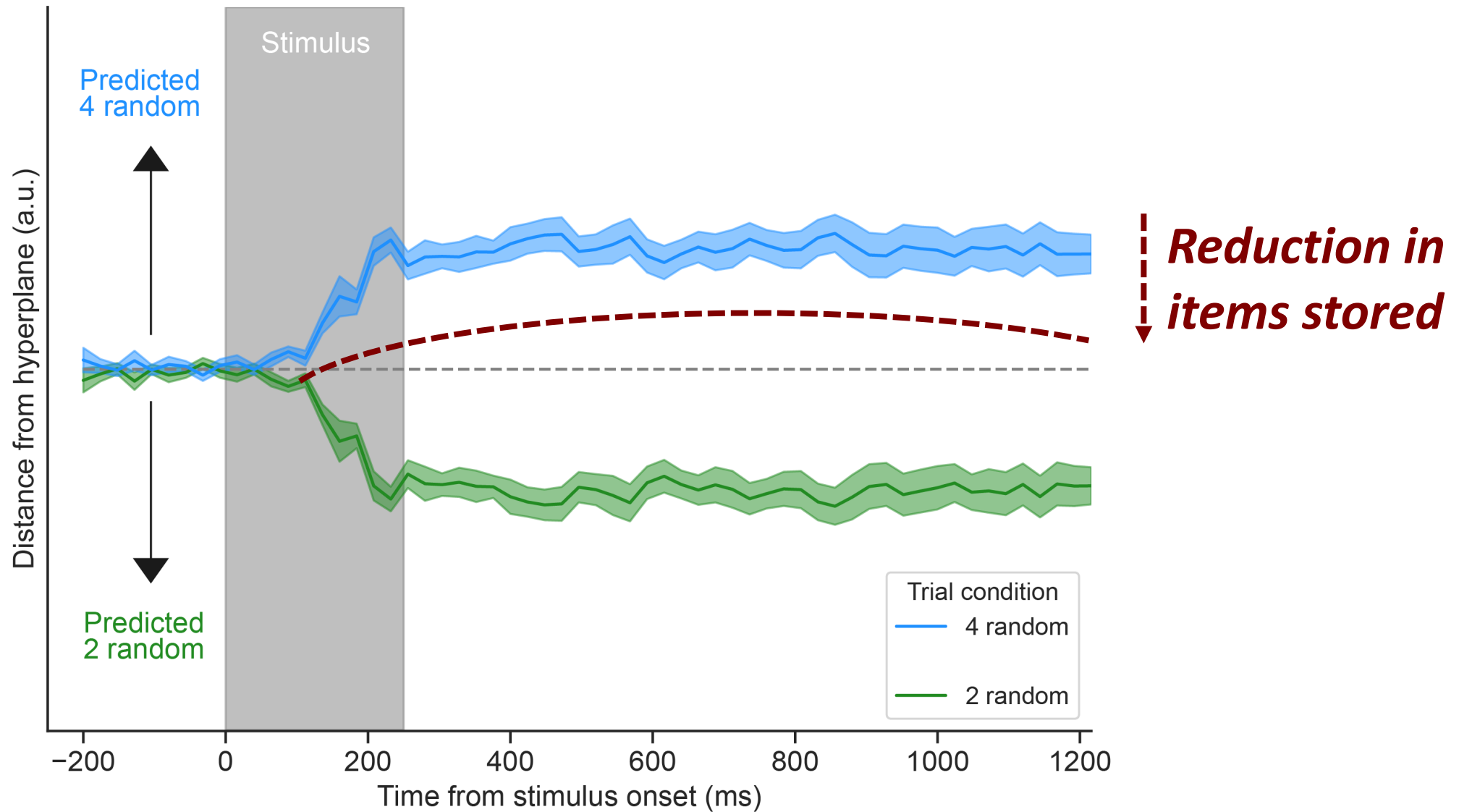
Train 2 random versus 4 random



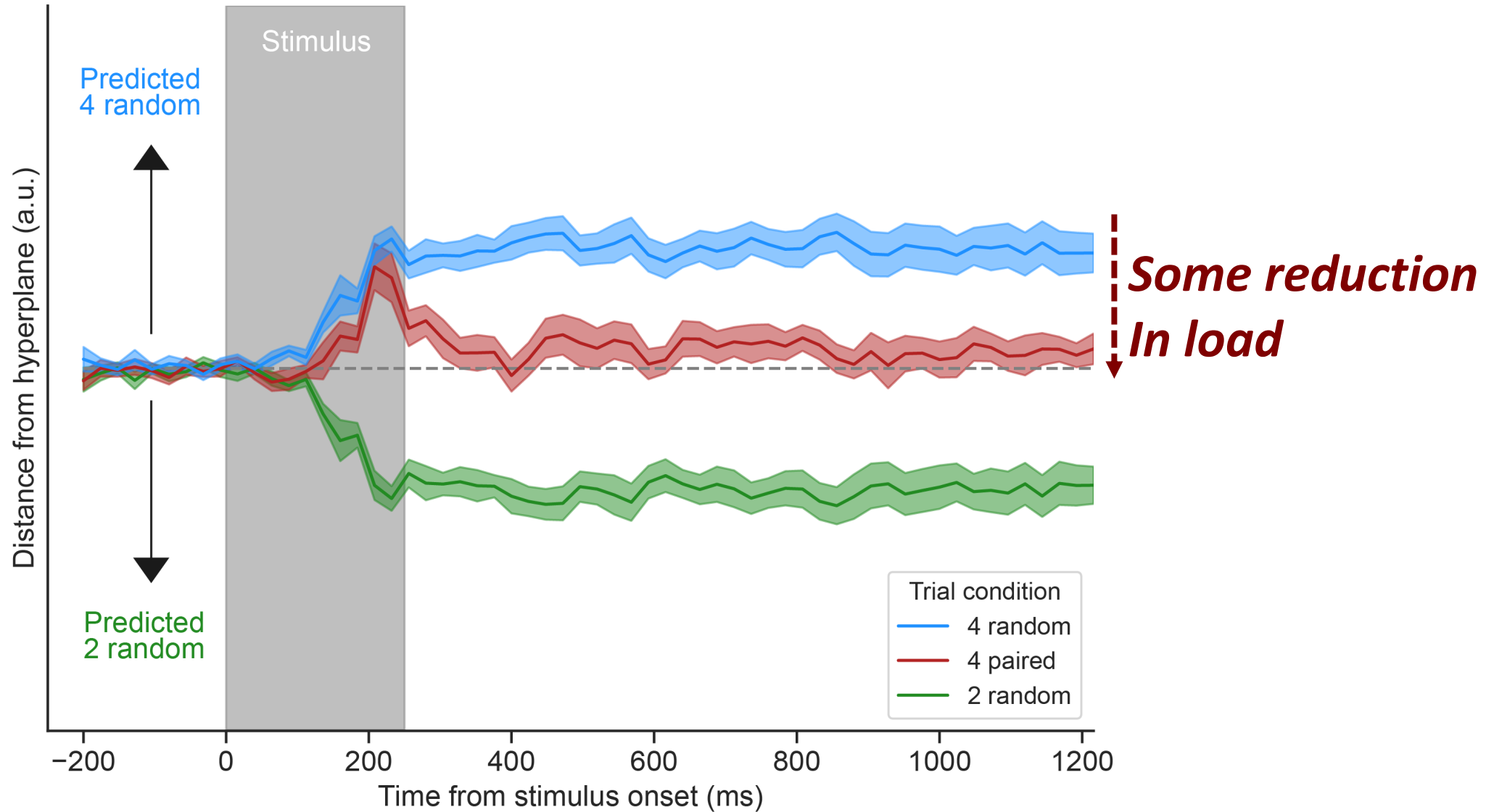
Train 2 random versus 4 random, where is 4 paired?



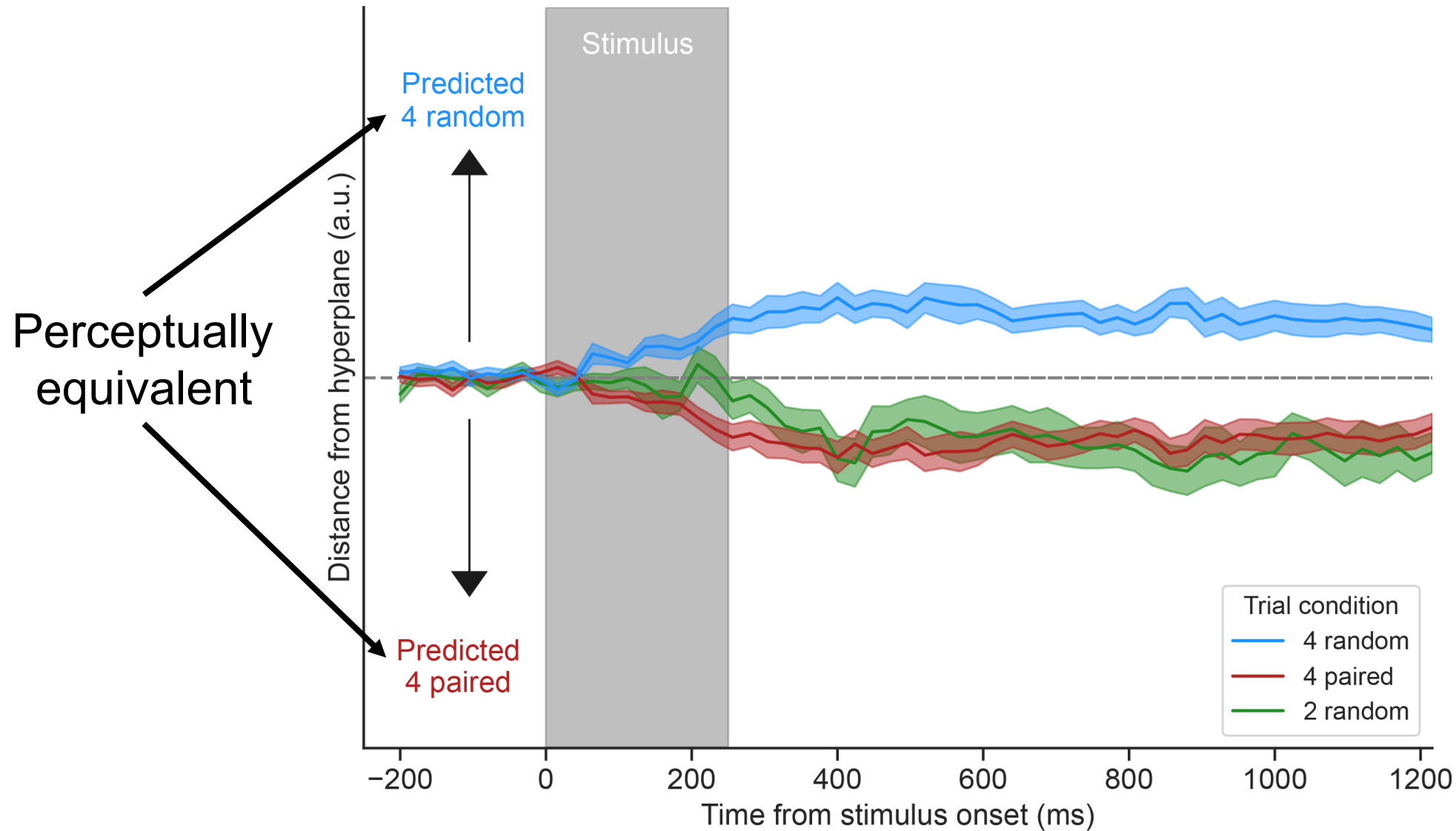
Train 2 random versus 4 random, where is 4 paired?



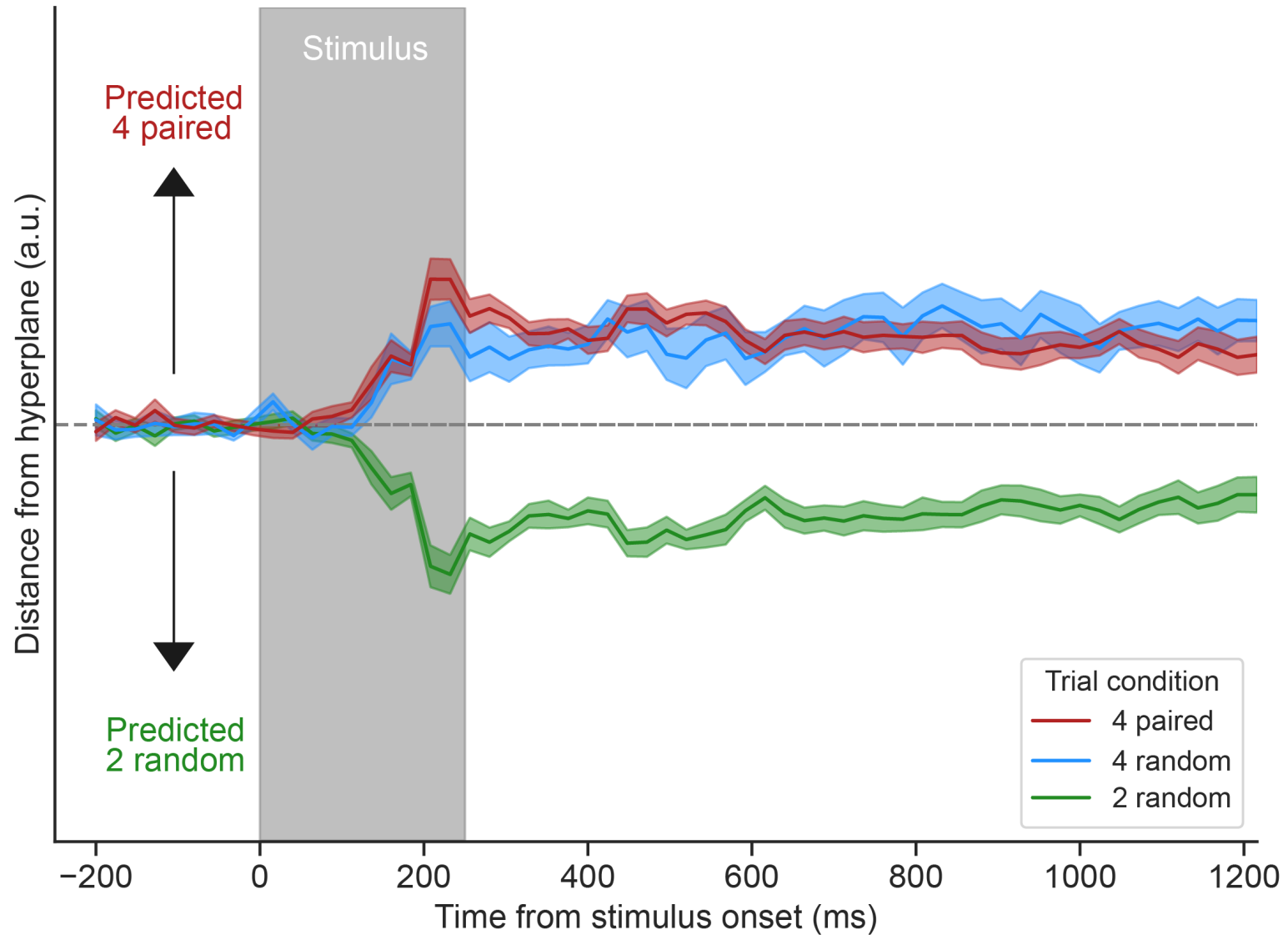
Train 2 random versus 4 random, test 4 paired



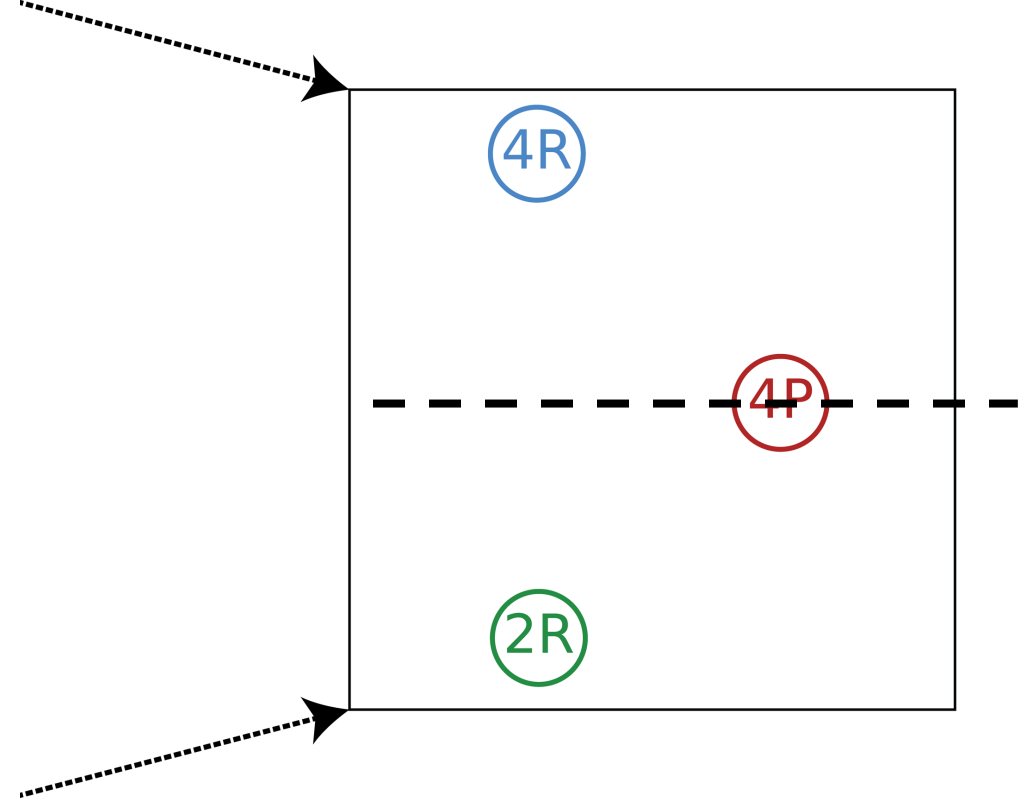
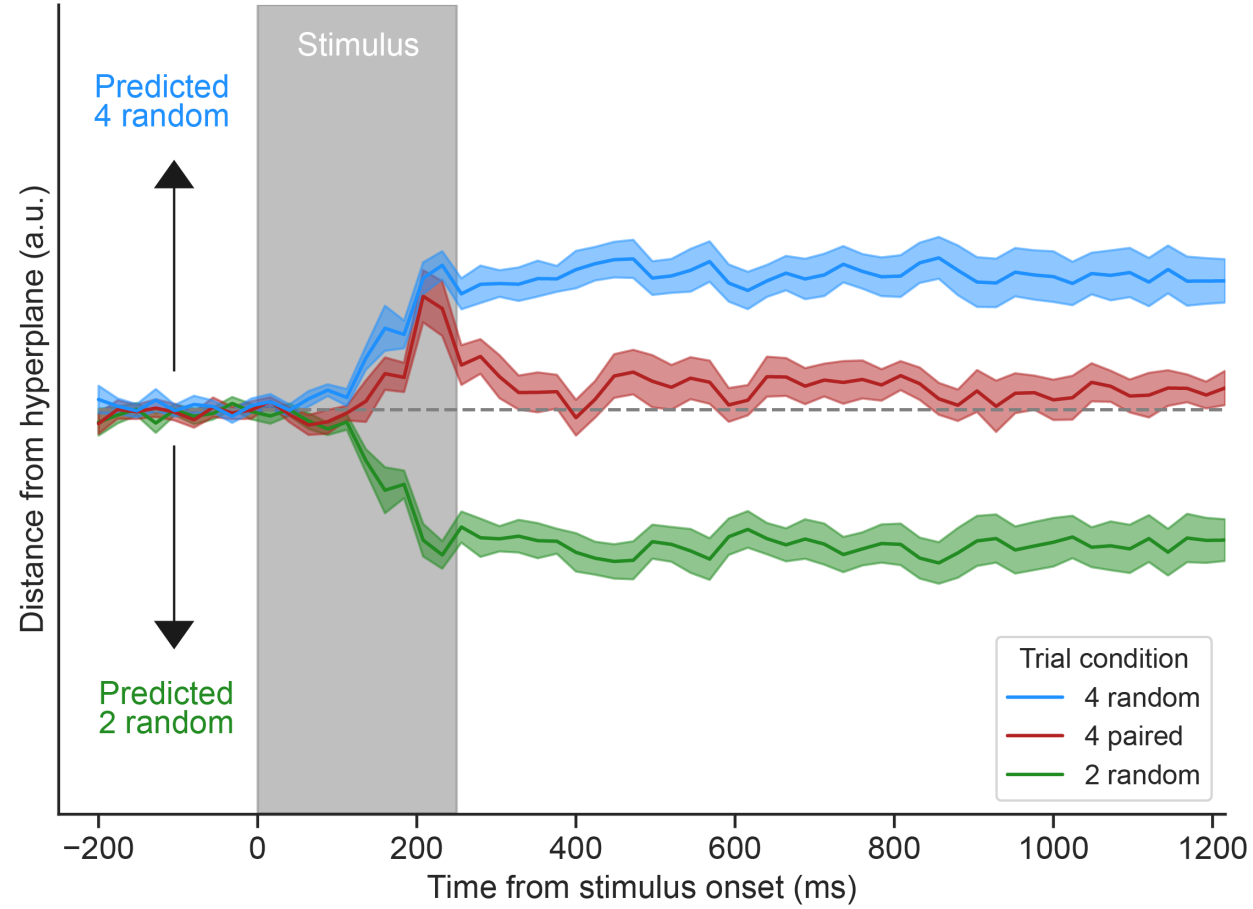
Train 4 random versus 4 paired, test 2 random



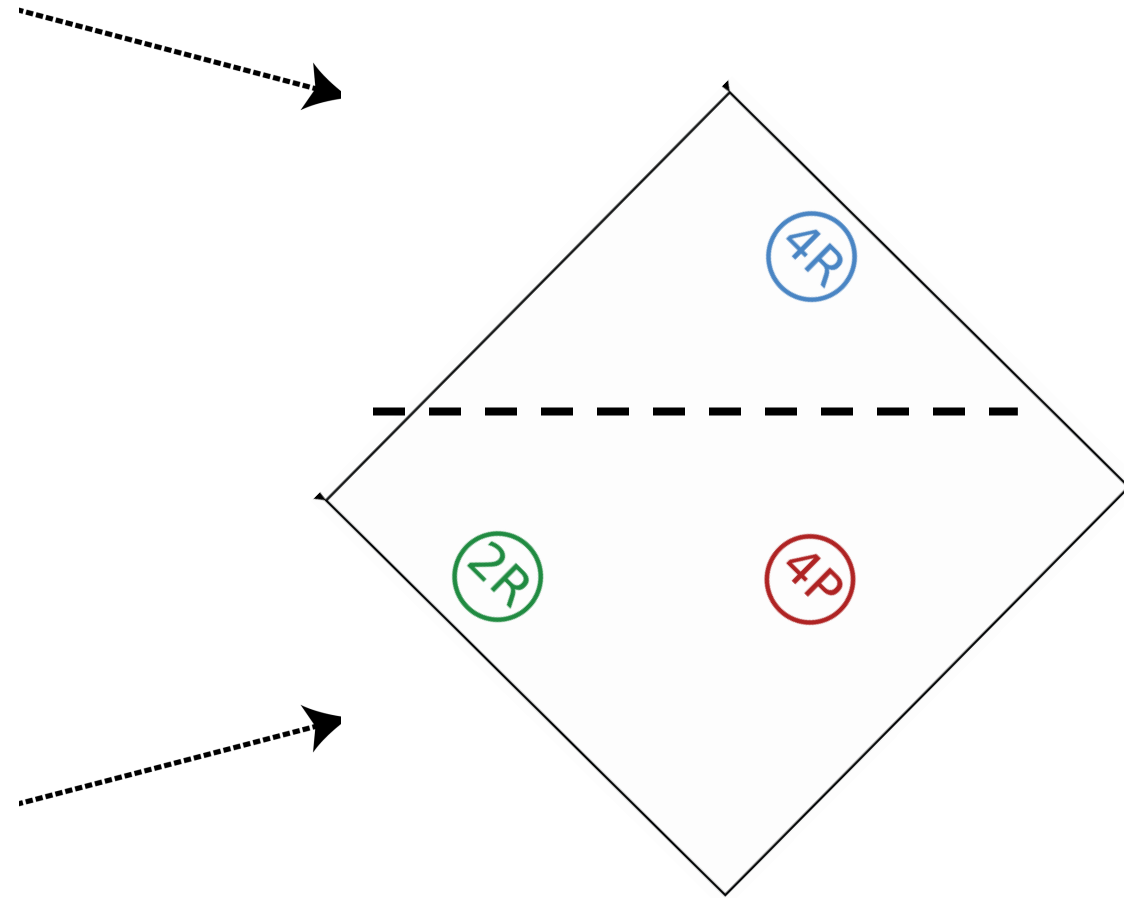
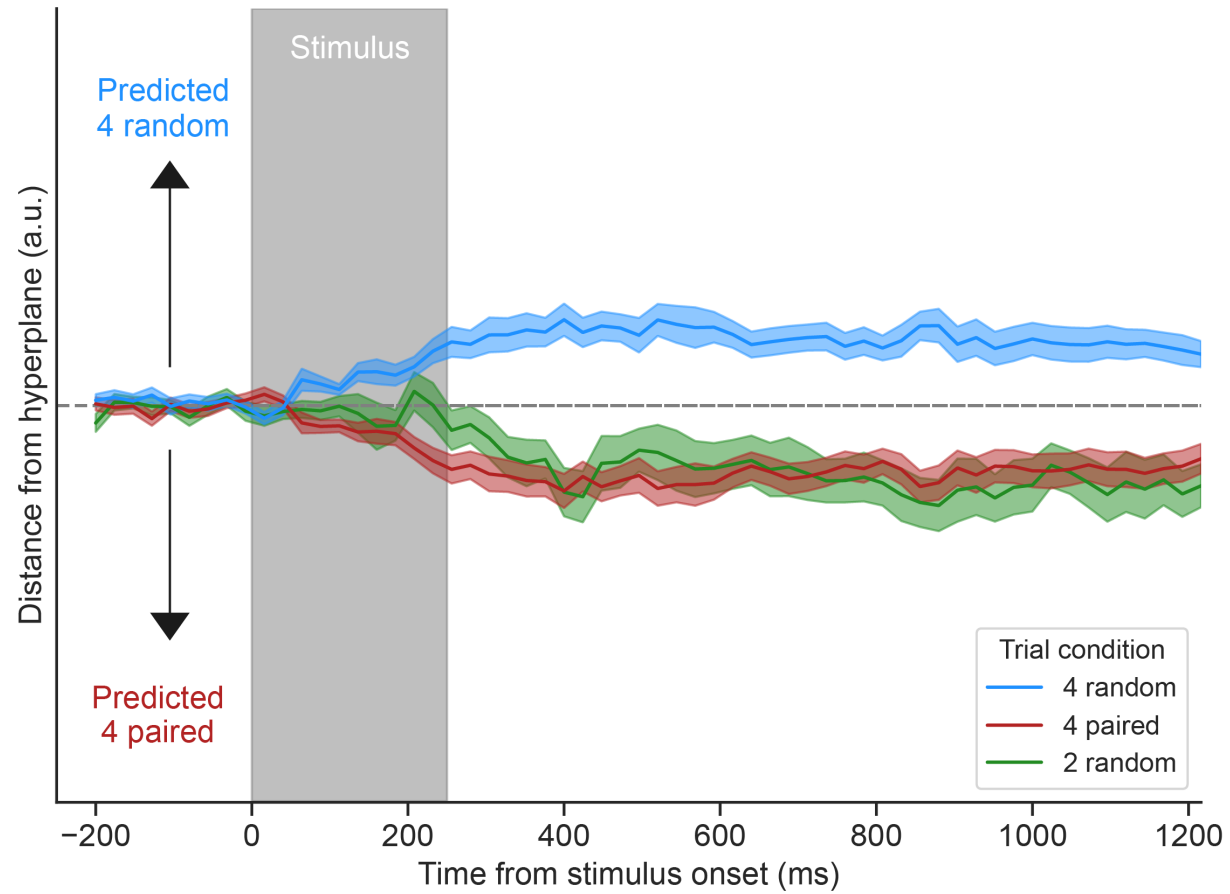
Train 2 random versus 4 paired, test 4 random



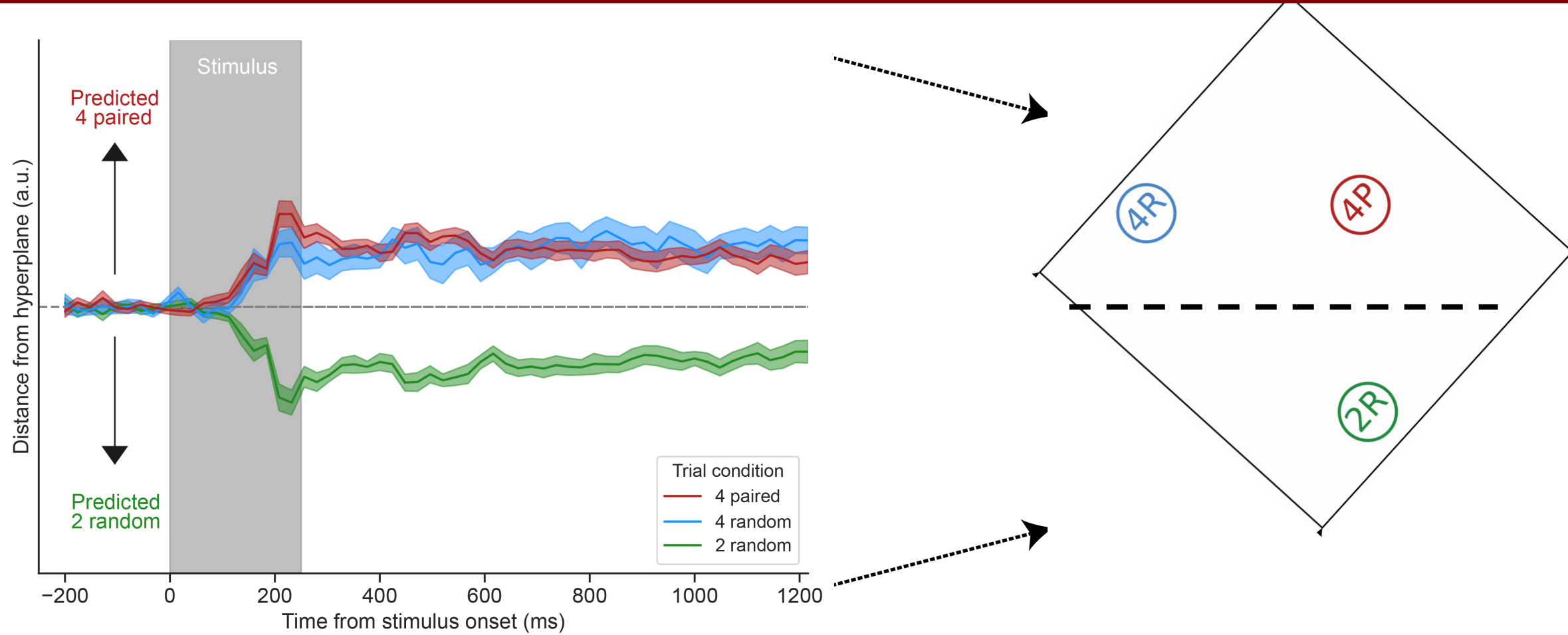
Multidimensional scaling



Multidimensional scaling

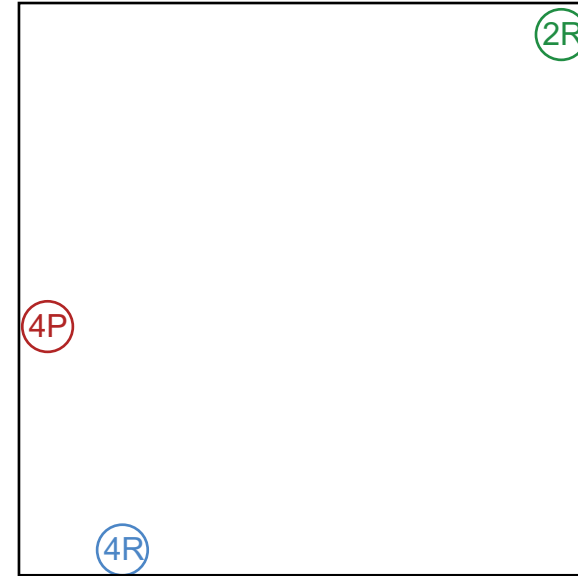
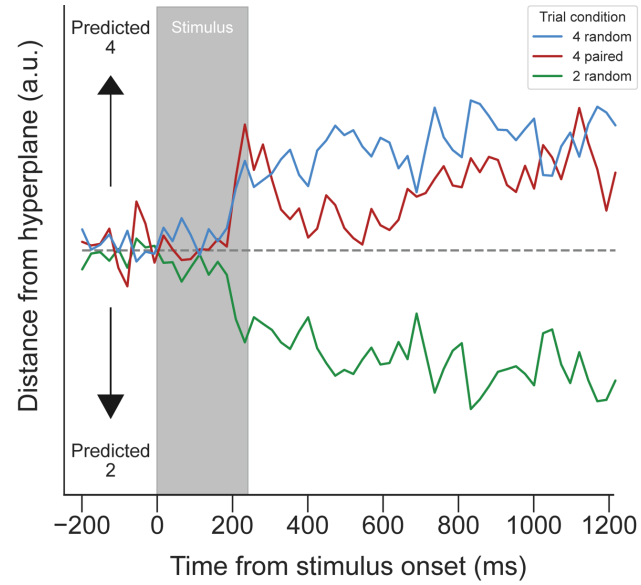


Multidimensional scaling

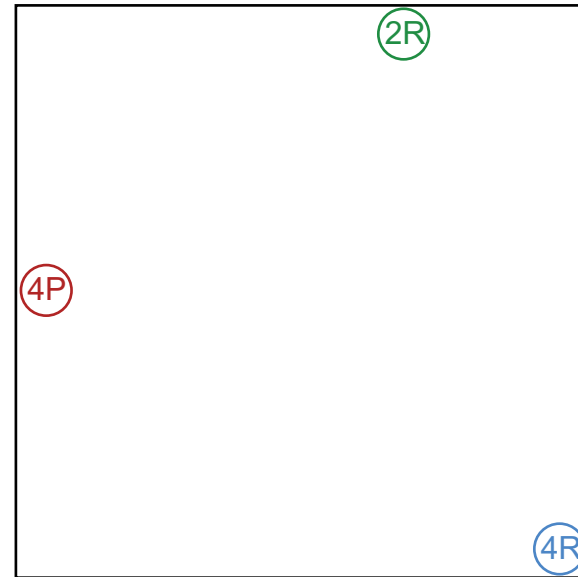
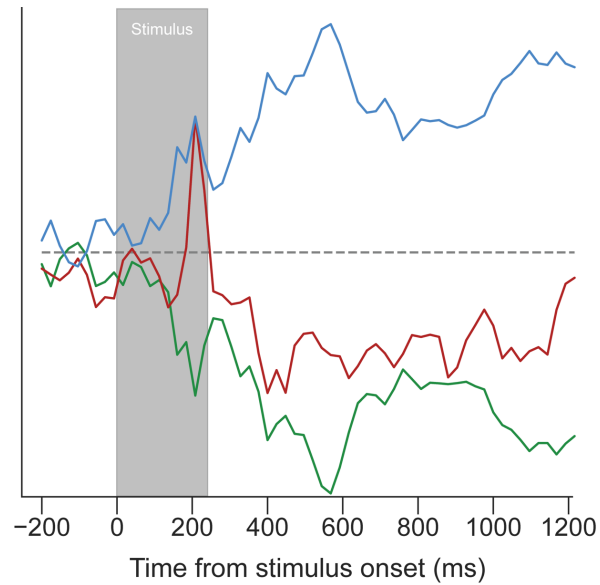


Individual differences

“Weak chunking”

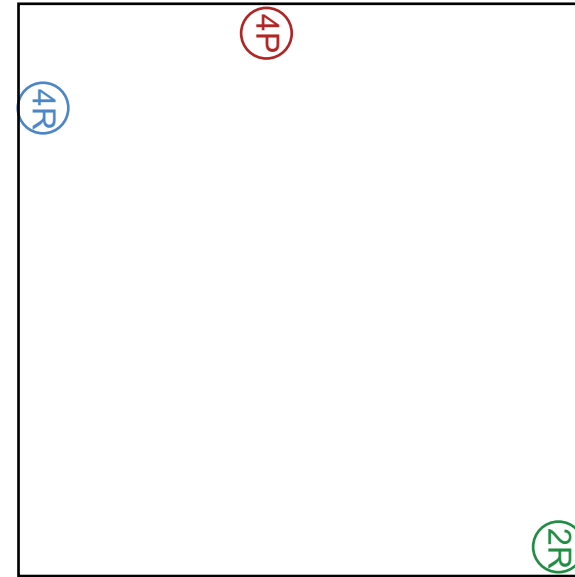
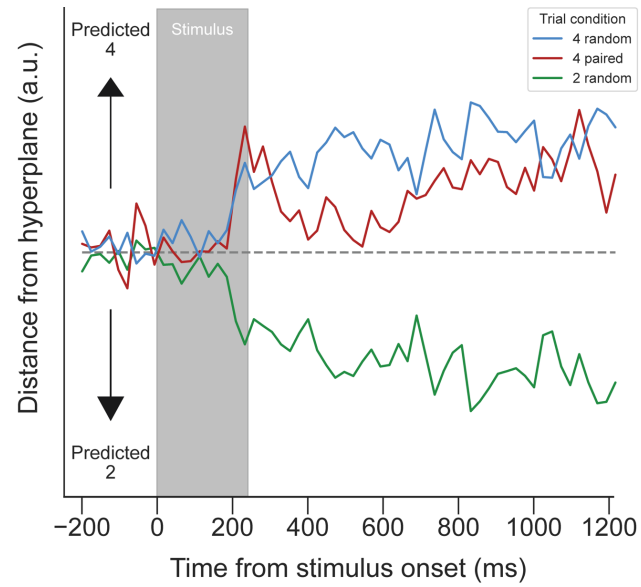


“Strong chunking”

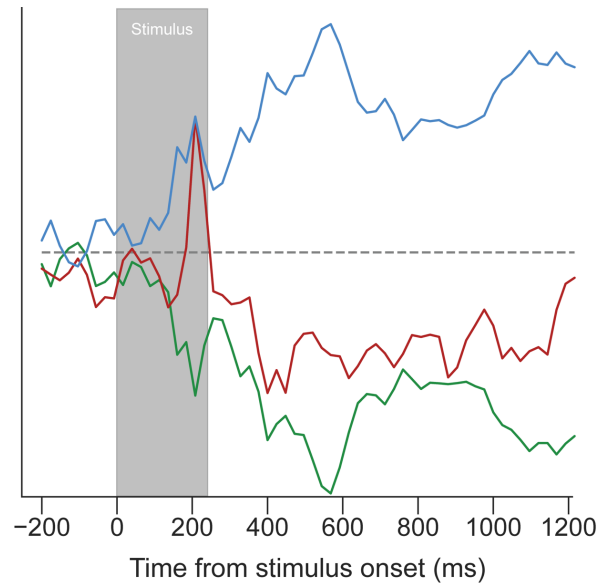


Individual differences

“Weak chunking”

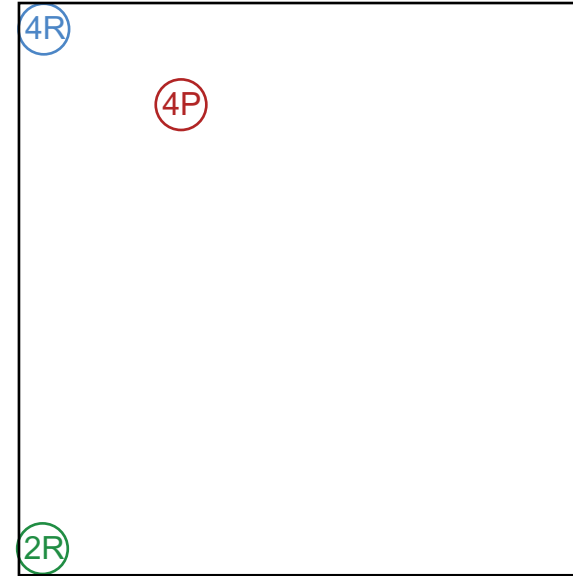
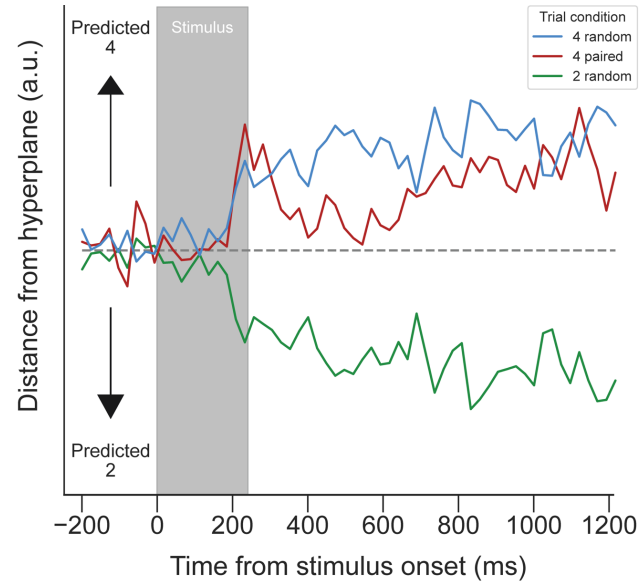


“Strong chunking”

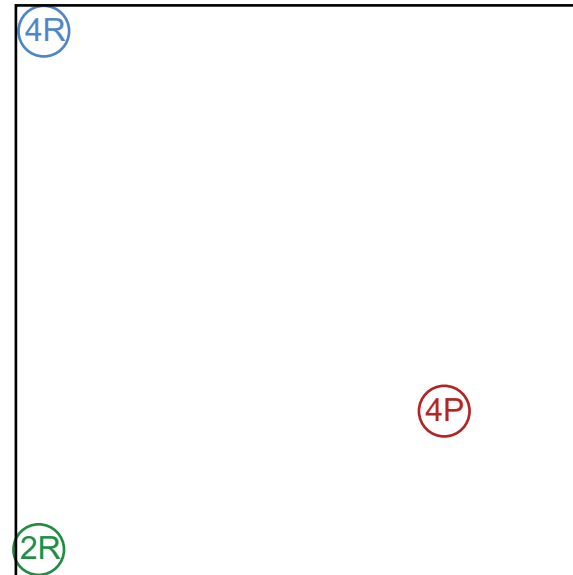
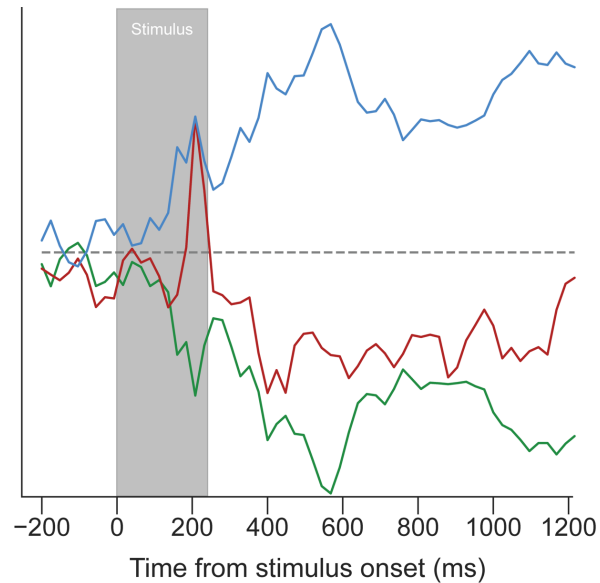


Individual differences

“Weak chunking”

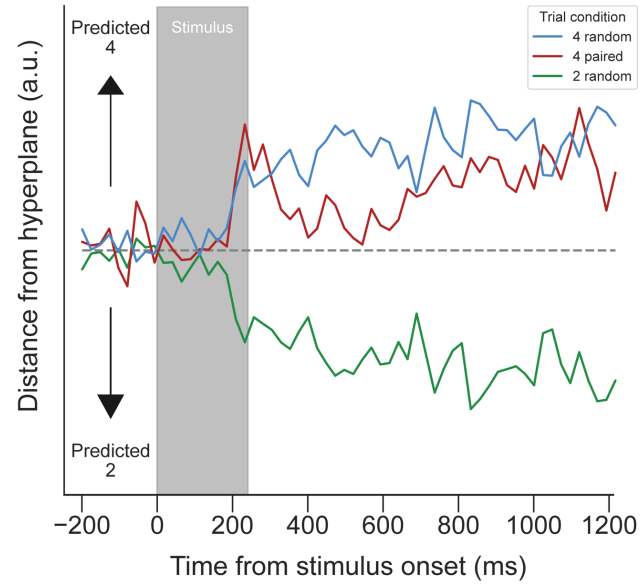


“Strong chunking”

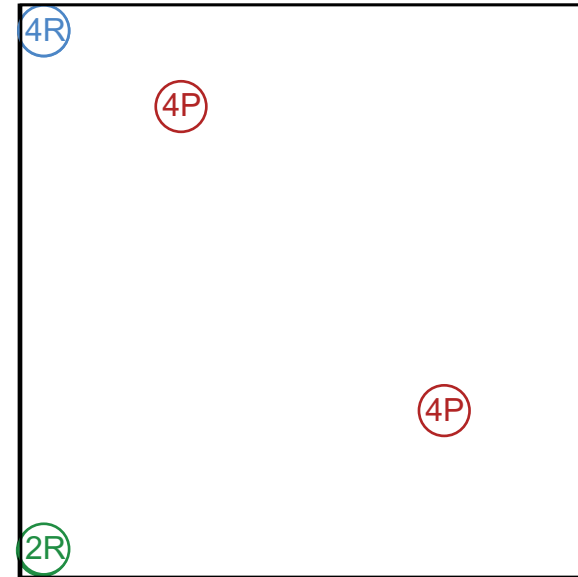
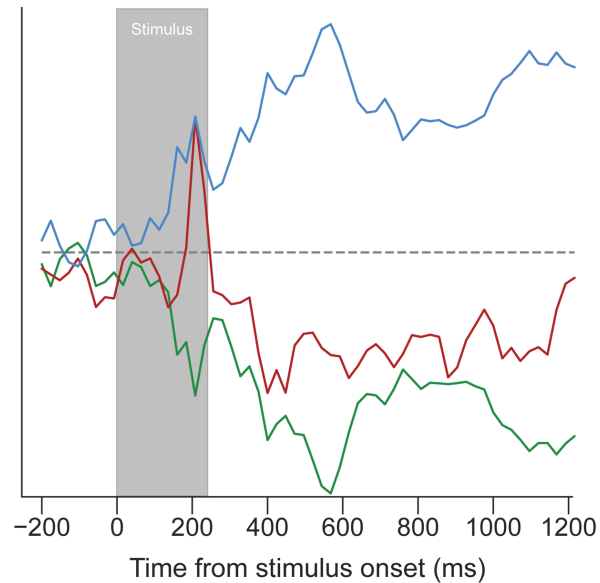


Individual differences

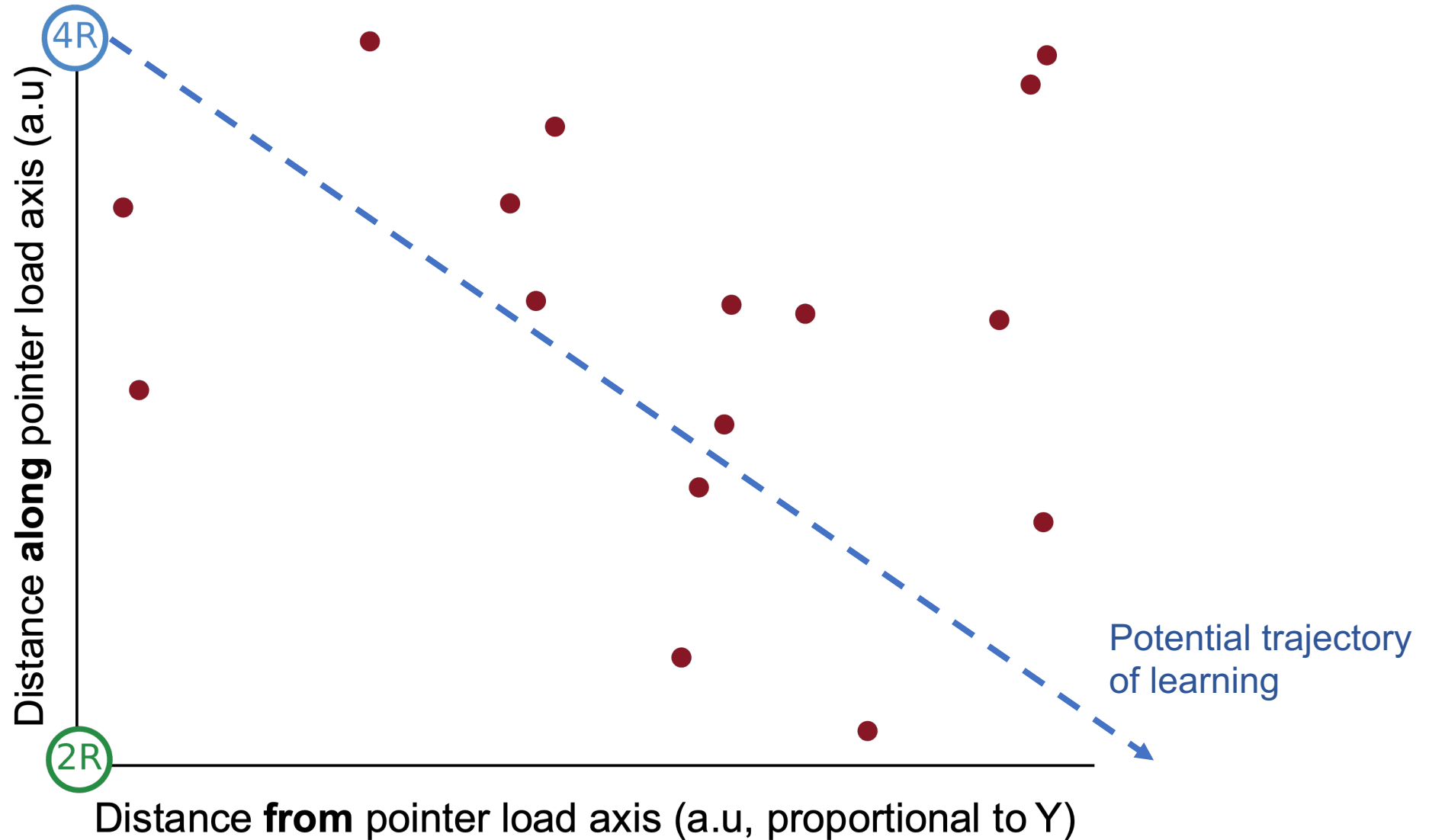
“Weak chunking”



“Strong chunking”



Individual differences



Experiment 2 - Training

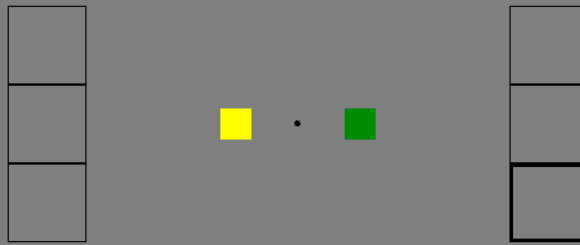
- Trained subjects to learn three color triplets



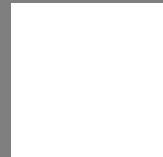
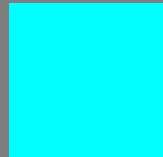
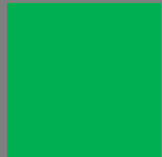
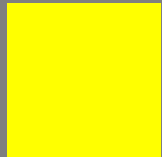
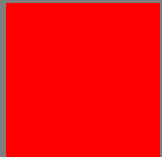
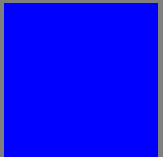
Experiment 2 - Training



Experiment 2 - Training

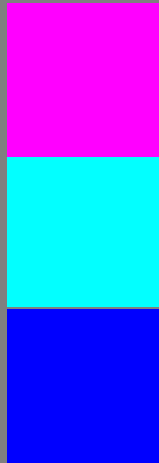


Awareness Test



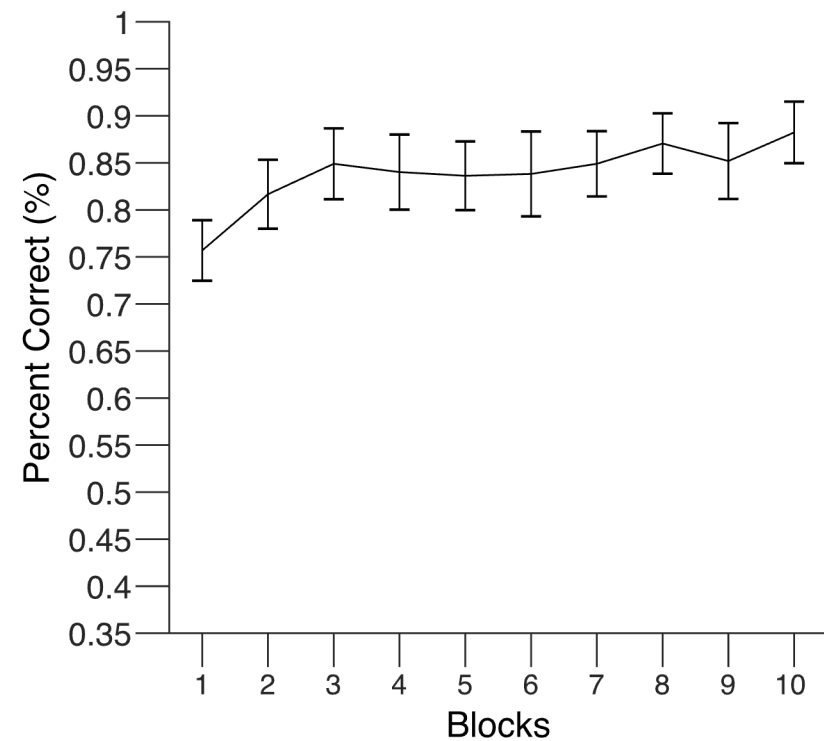
Awareness Test

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

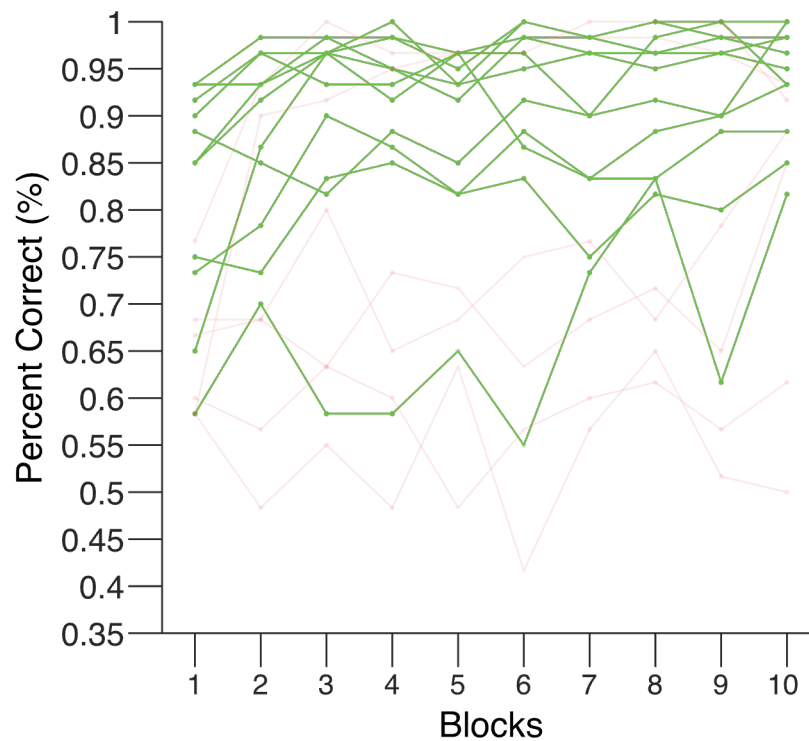


Experiment 2 Training

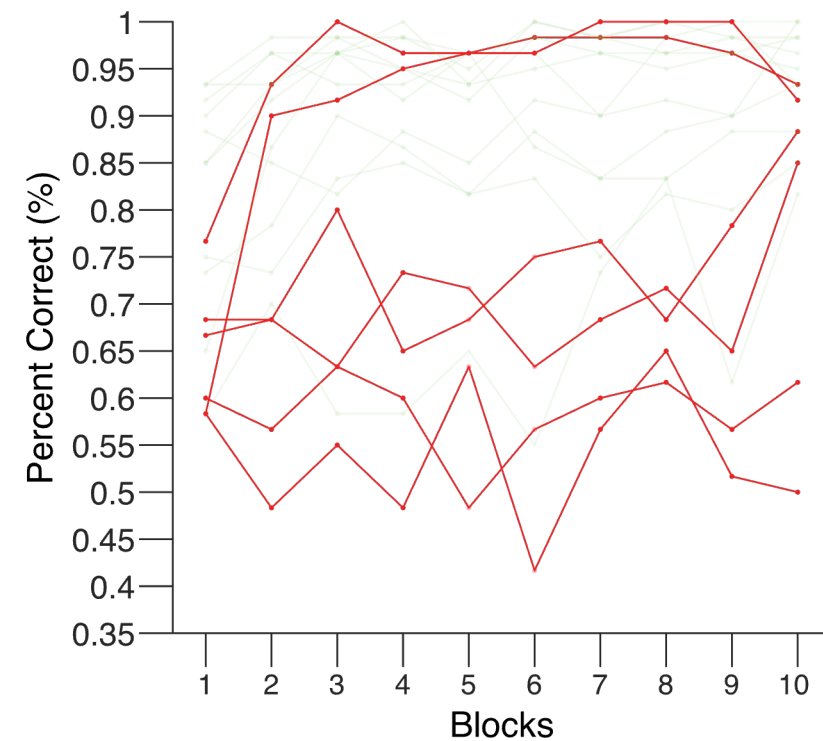
Aggregate



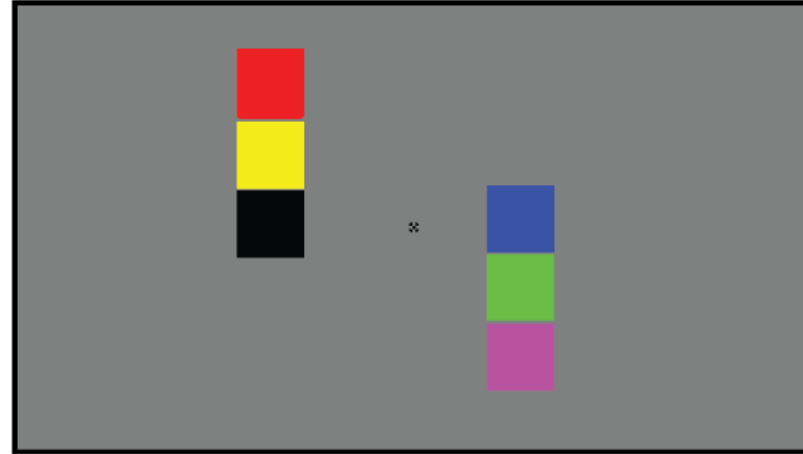
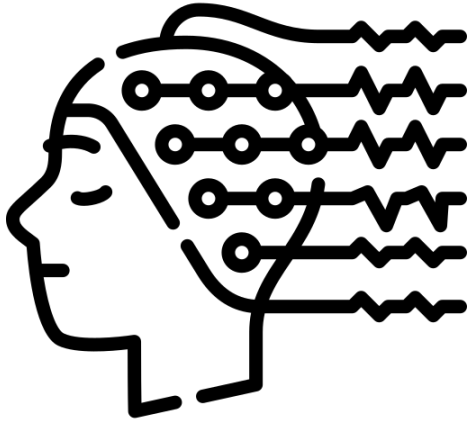
Learners



Non-learners

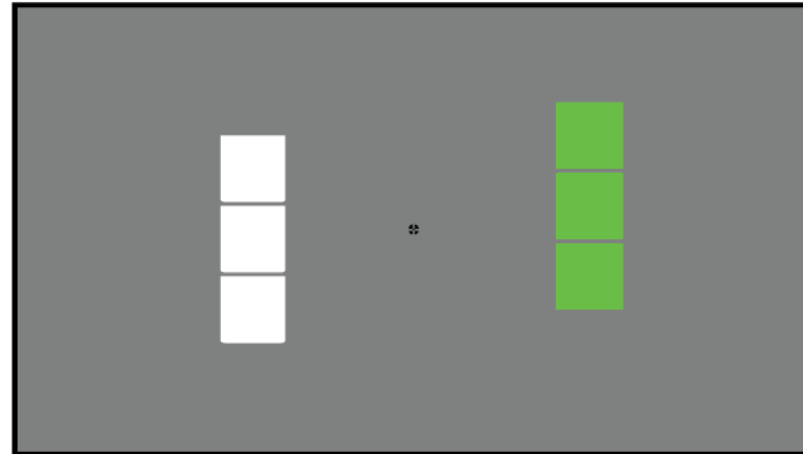


Experiment 2: EEG session



Six random
Six chunked

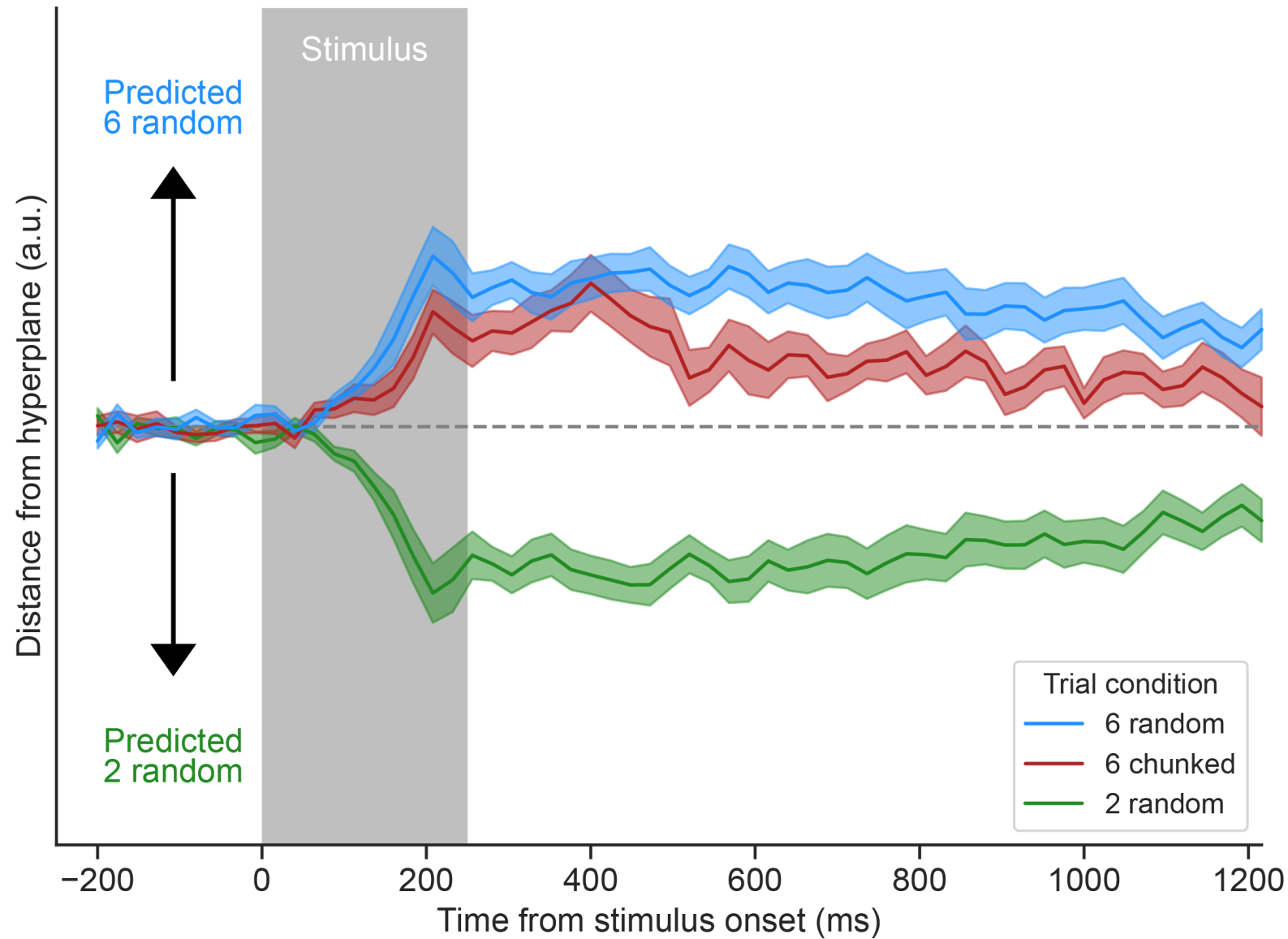
Perceptually
equivalent



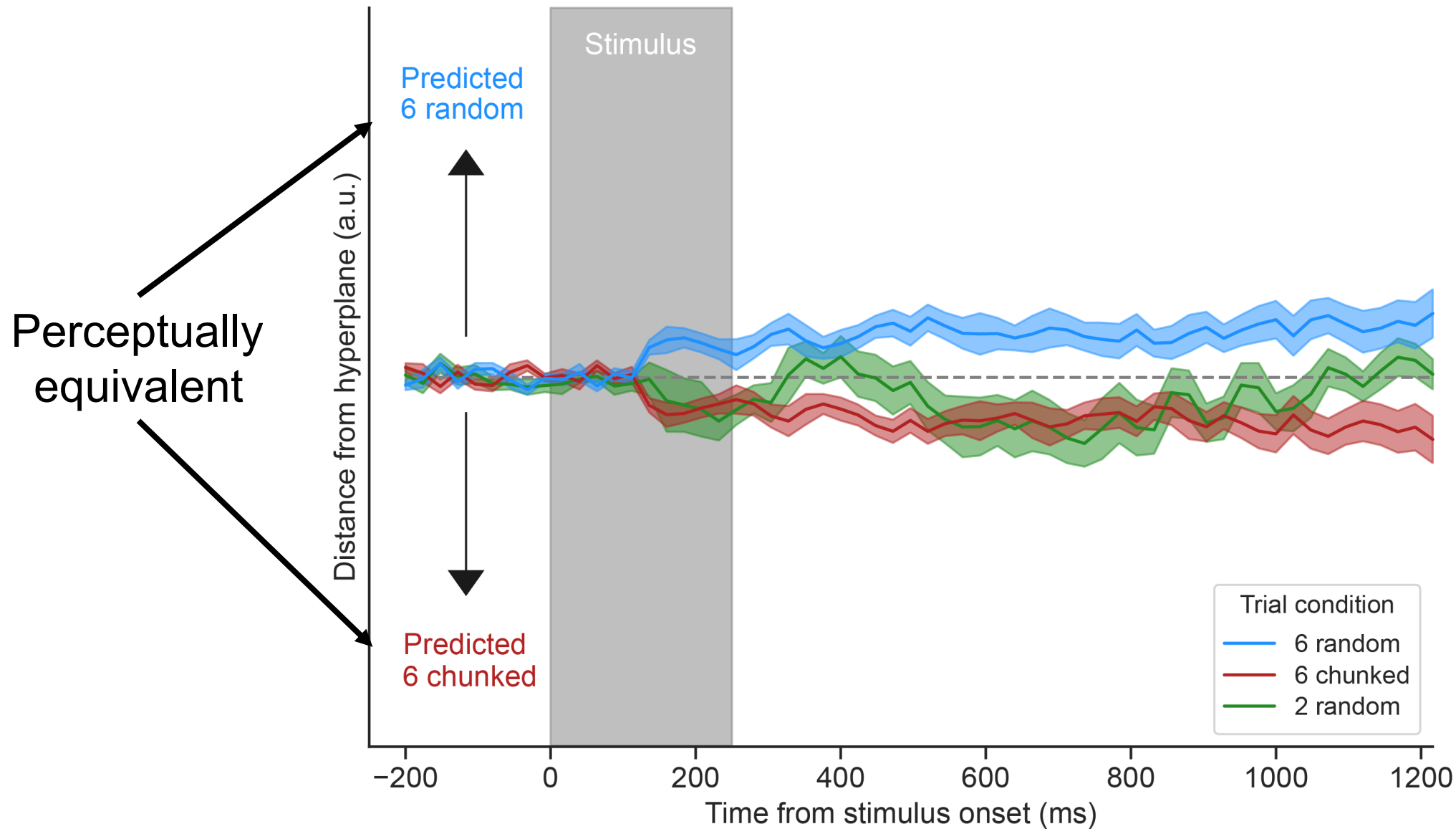
Two random



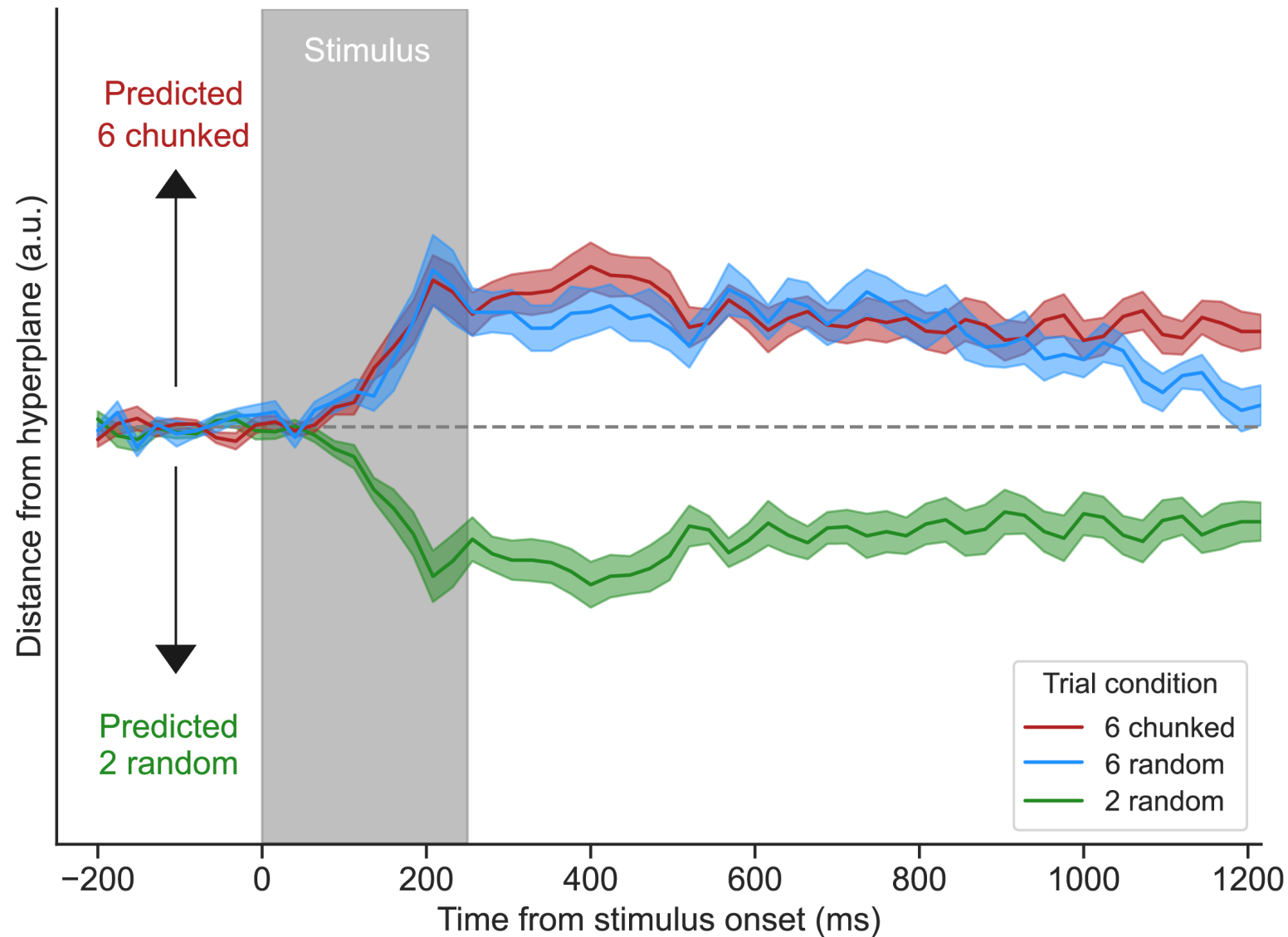
Train 6 random versus 2 random, test 6 chunked



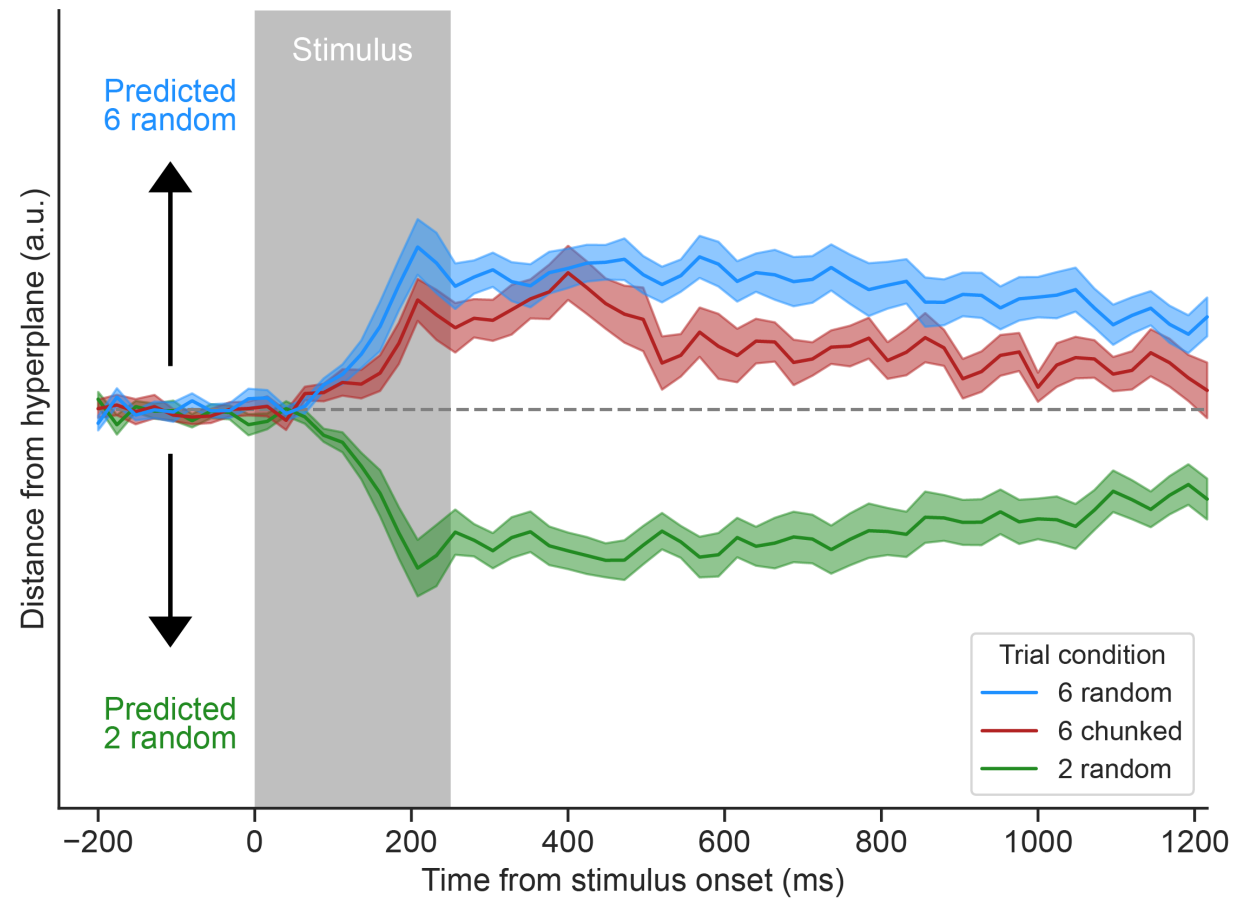
Train 6 random versus 6 chunked, test 2 random



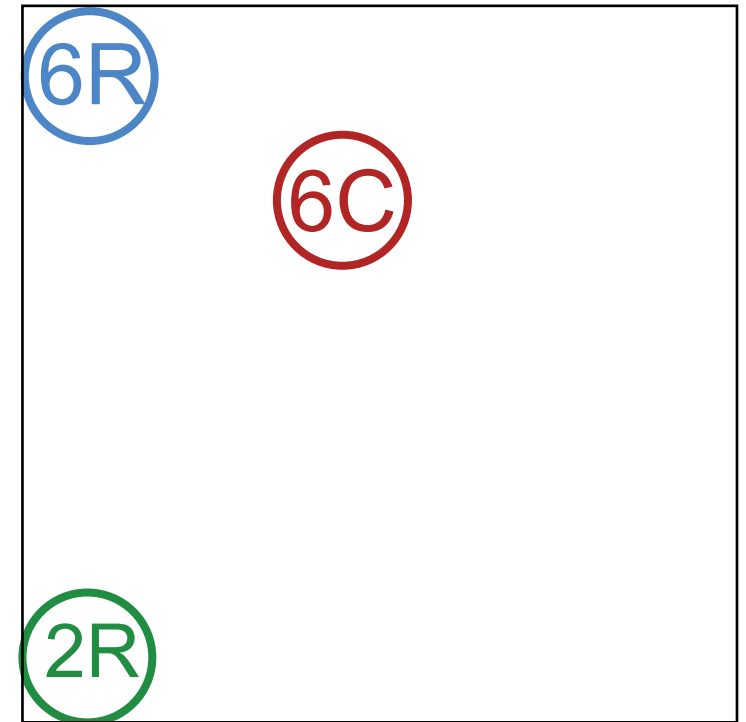
Train 2 random versus 6 chunked, test 6 random



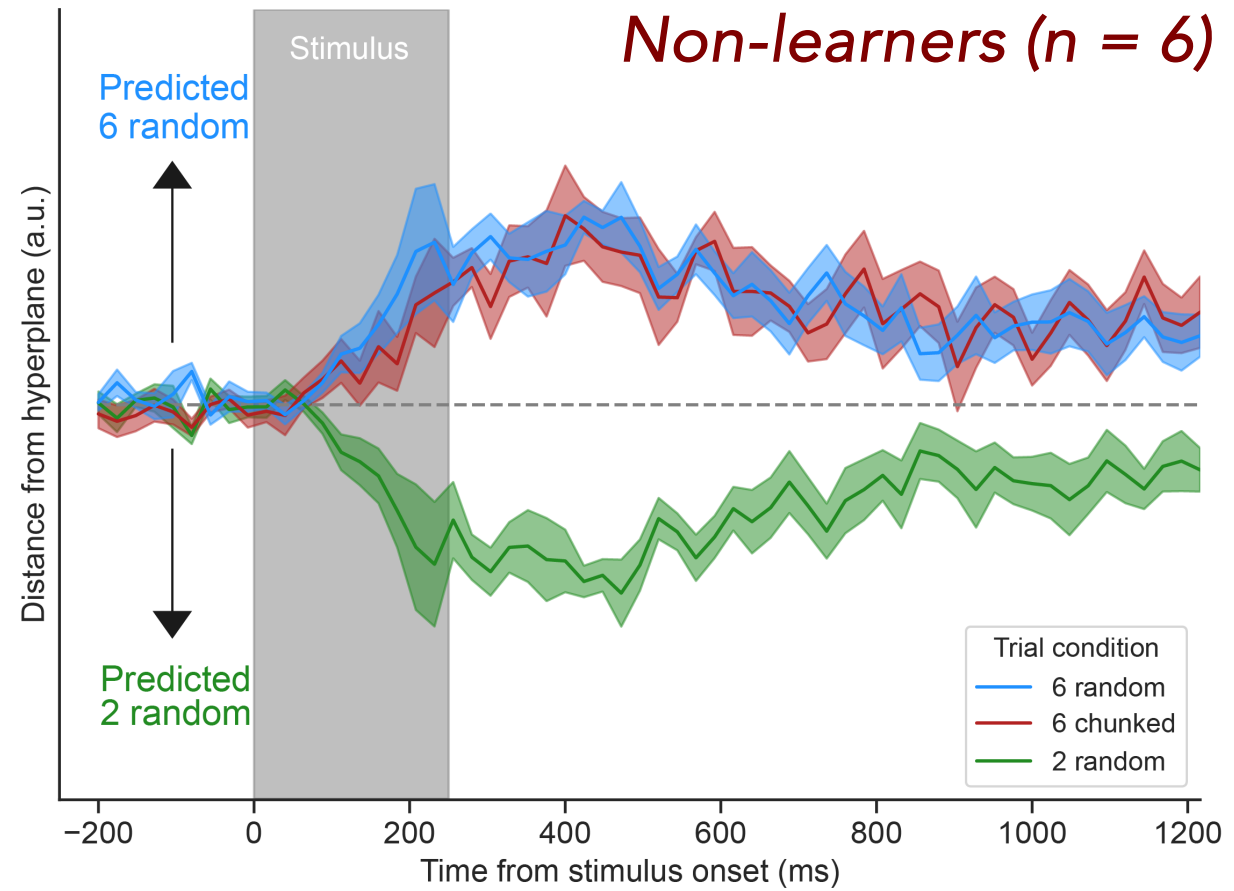
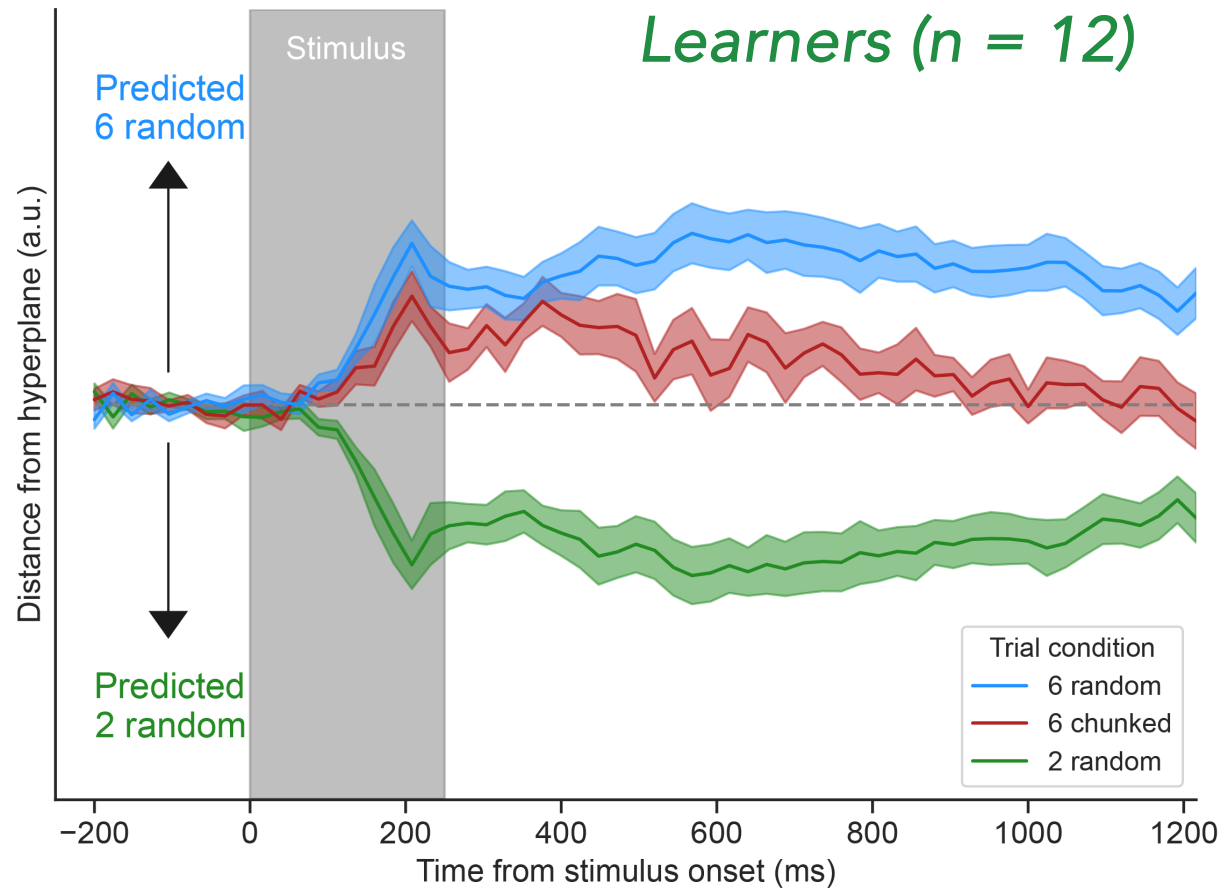
Train 6 random versus 2 random, test 6 chunked



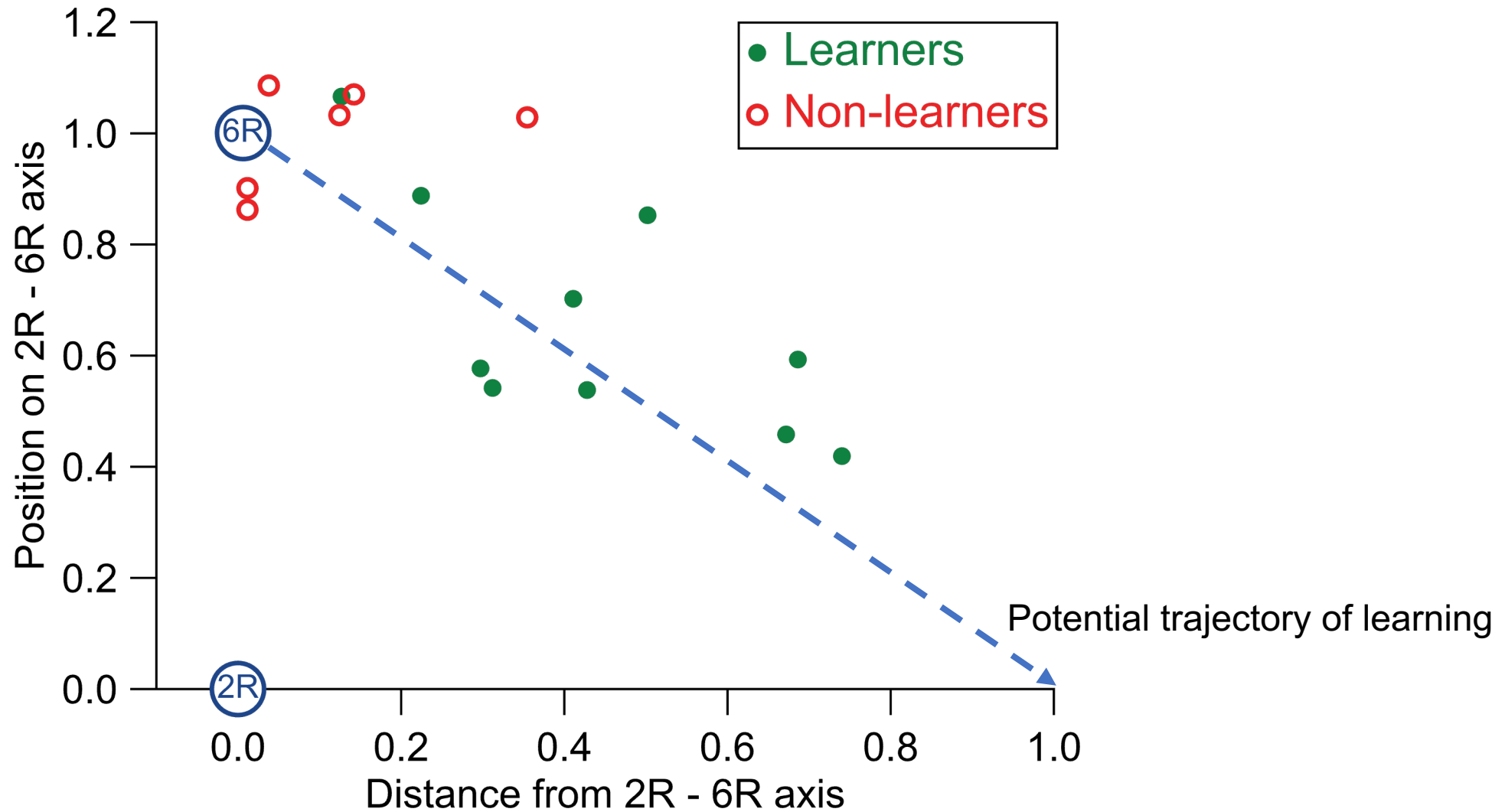
Multidimensional scaling



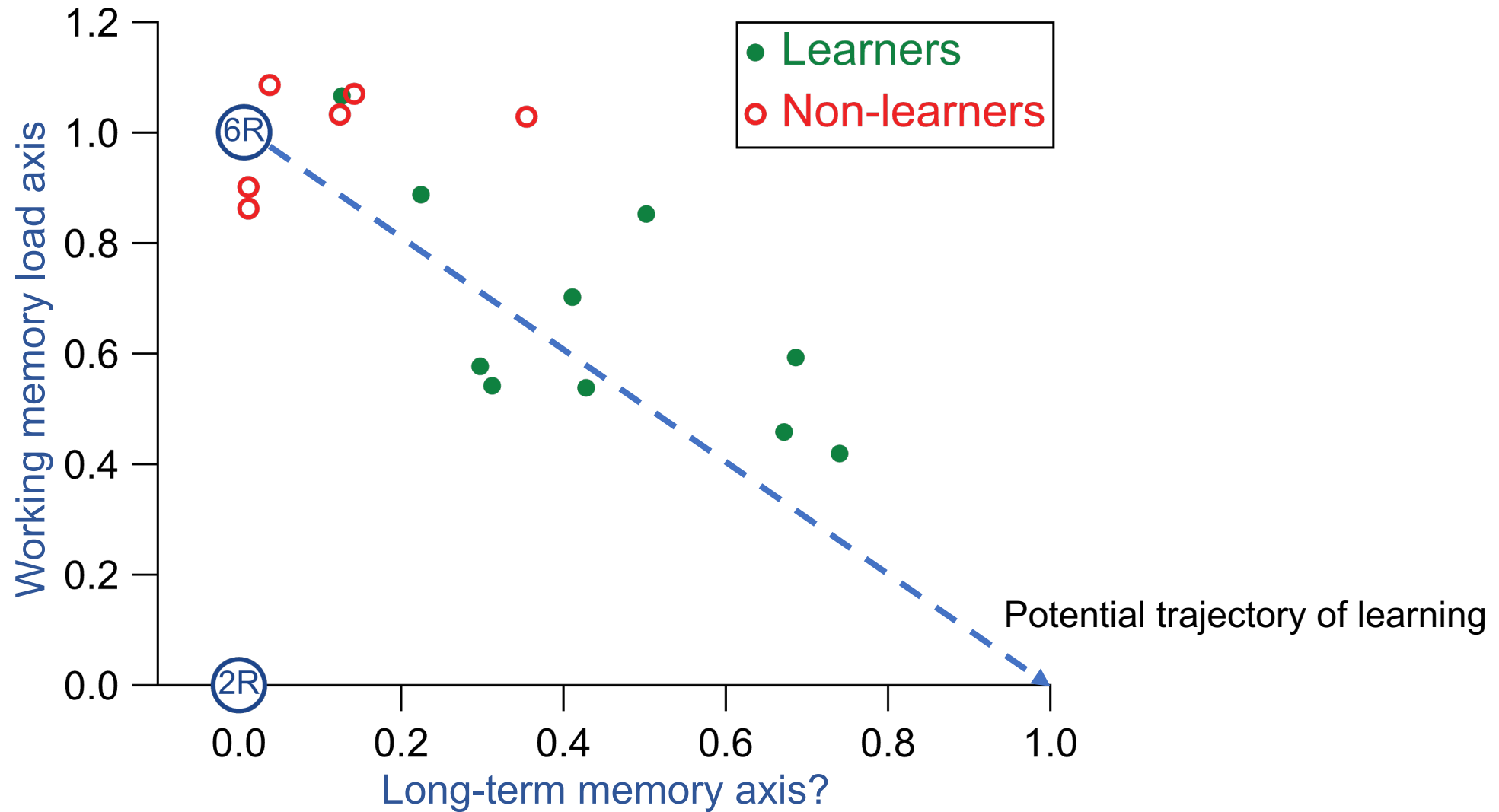
Learners vs non-learners



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



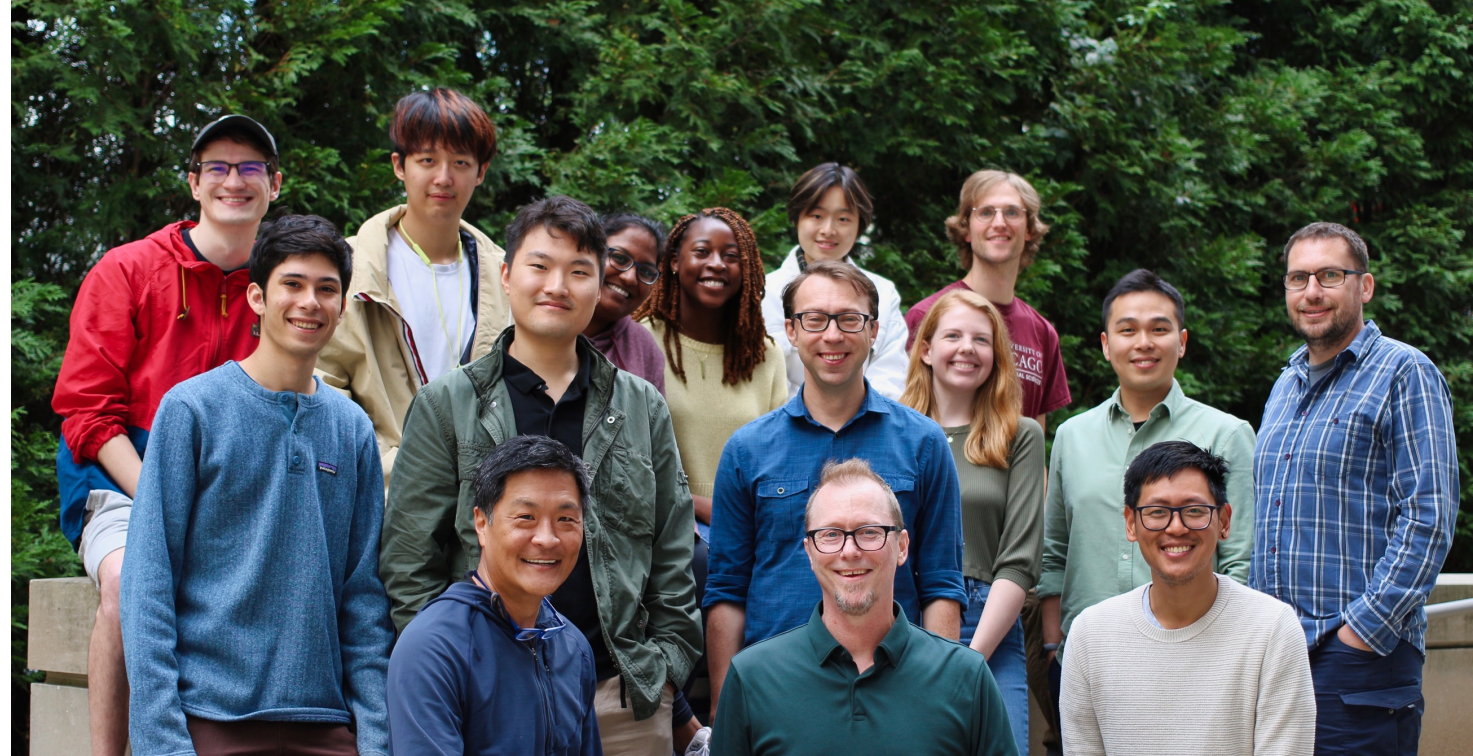
William Thyer



Edward Awh



Leo Chang



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



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



Follow me:



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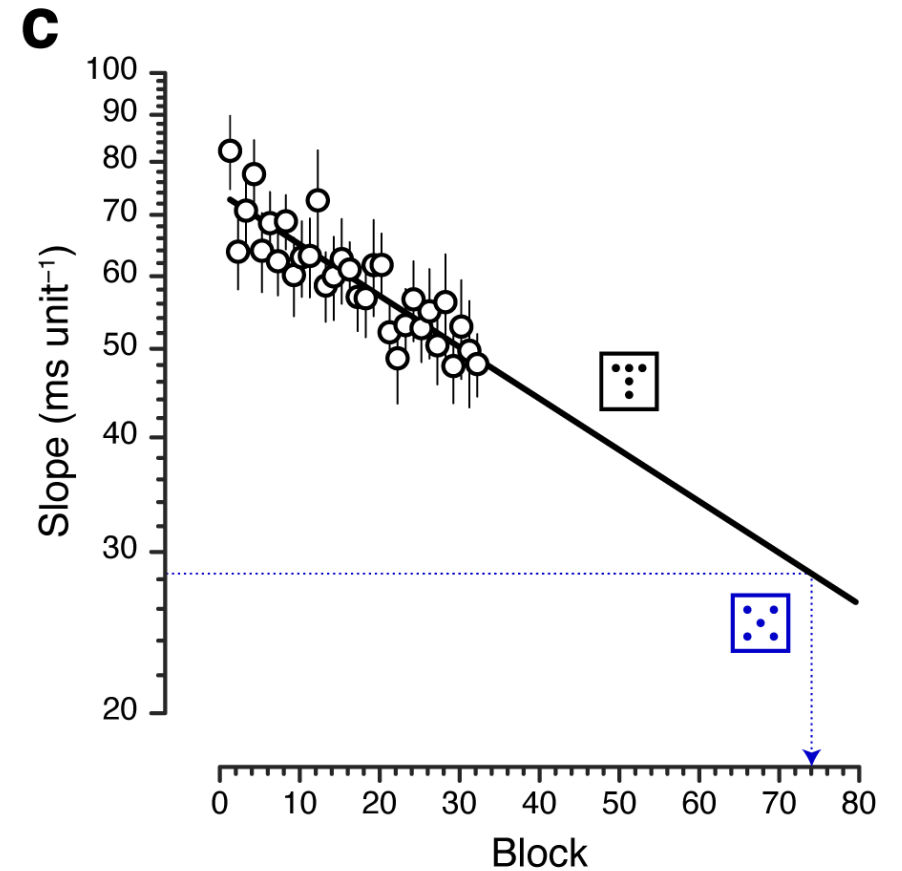
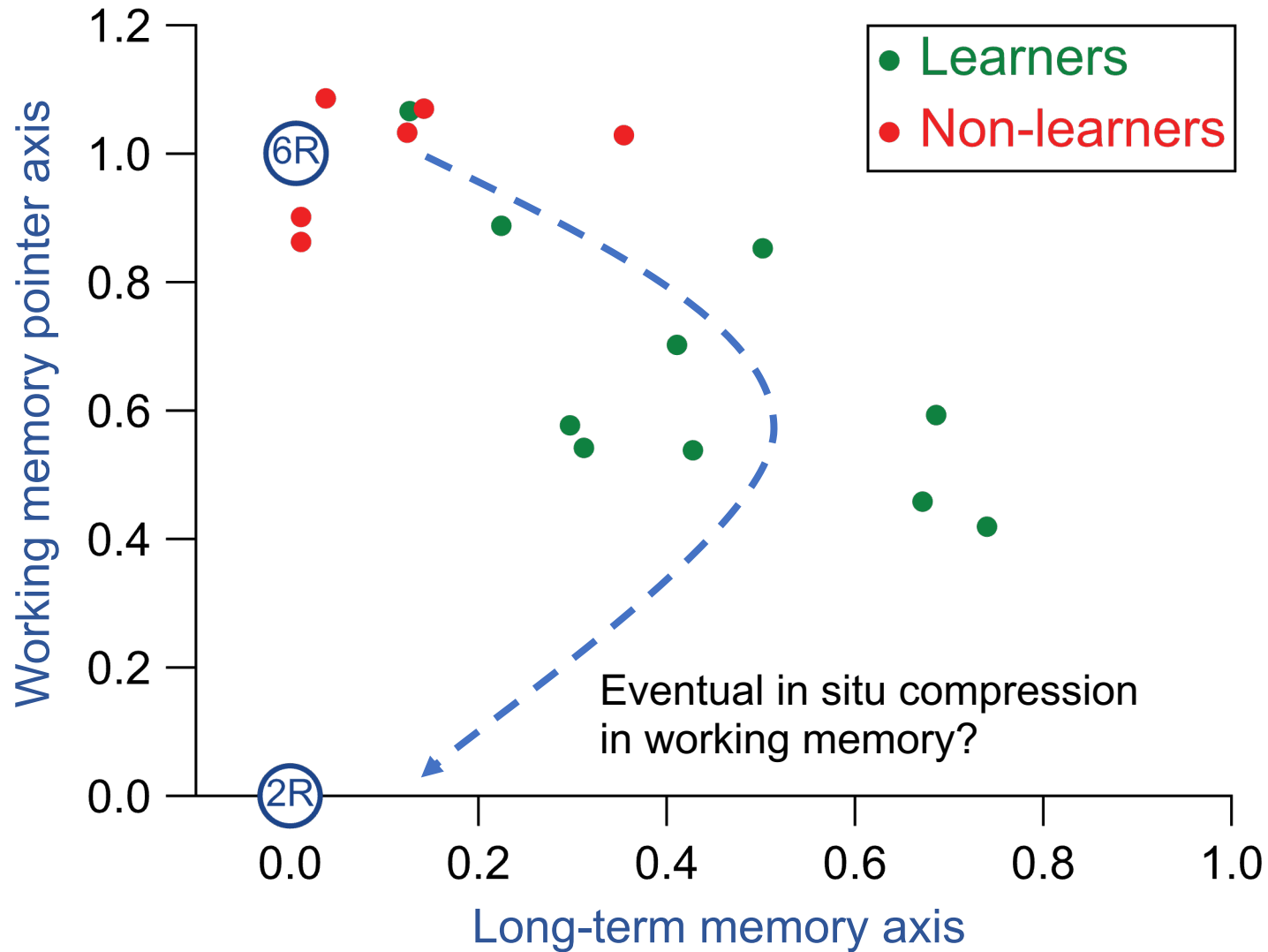


Bluesky

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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)

