

# The Role of Brain Oscillations in Deep Predictive Learning: Behavioral Approaches

Maryam Zolfaghar

**Supervisor:** Prof. Randall O'Reilly

Department of  
Computer Science

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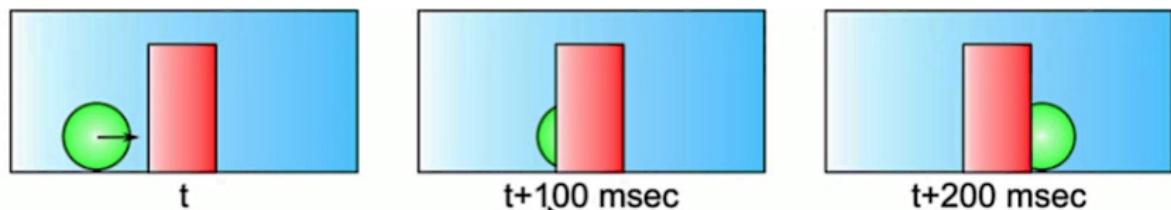
## Outline

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# Introduction

# Predictive Learning



## Key Points

- We are constantly predicting our future
- Learning signals (i.e., the prediction error) driven by the difference between the brain's top-down prediction and bottom-up real world information
- **Hypothesis:** Predictive learning occurs based on a temporal organization (alpha cycle, 100ms).

## Current Research

## Current Research

**What would be a good experimental paradigm to test  
predictive learning?**

## Experimental Design

# Experimental Design

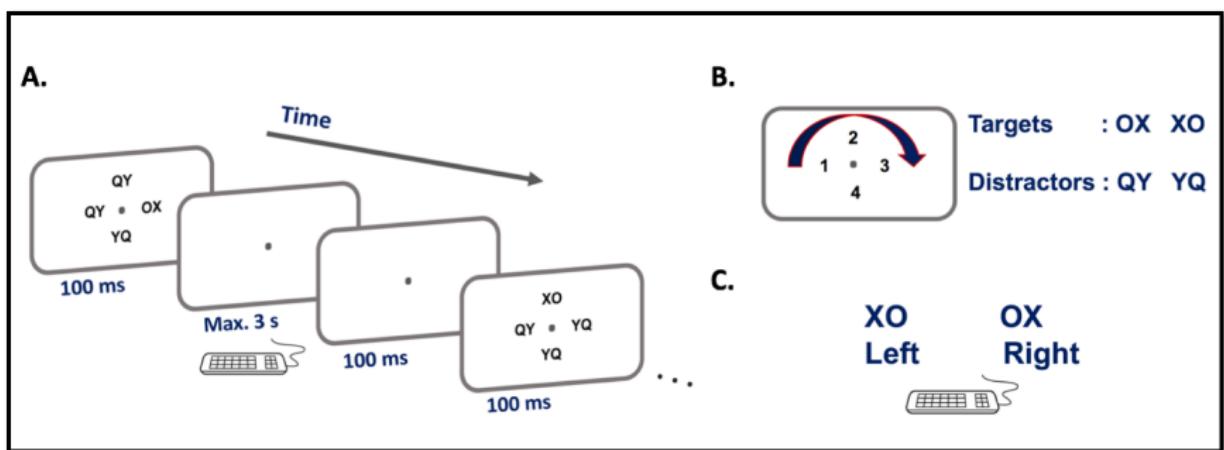
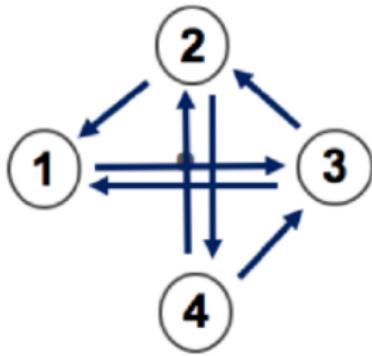


Figure: N. Deroost et. al, "Spatial processing and perceptual sequence learning in SRT tasks" (2006).

- Statistical-sequence learning
- Pure perceptual learning

## Experimental Design



S1: 42132431  
S2: 13423124

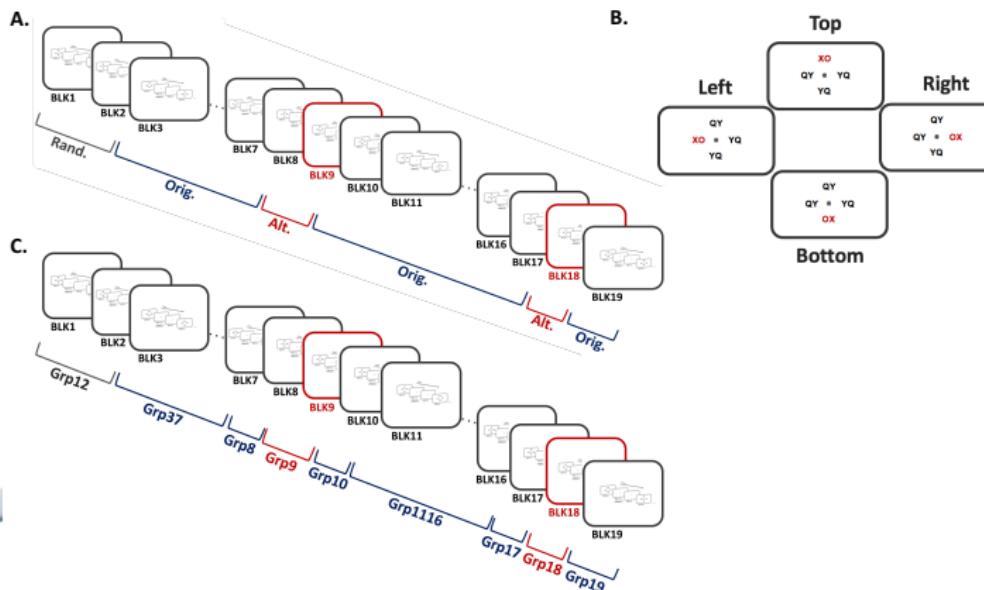
- **Implicit Learning**

# Experiment 1

## Procedure

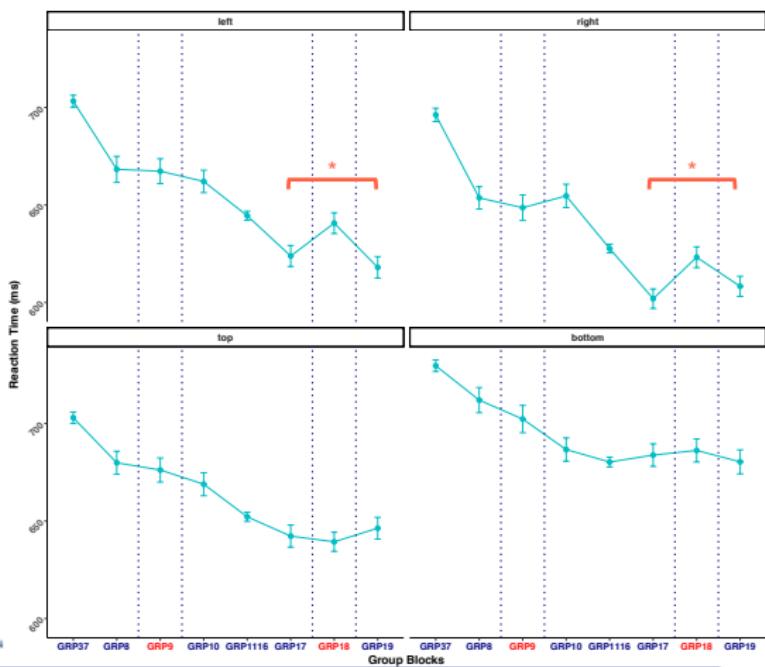
## Goals:

- Replicate the sequence learning without including any motor related learning



## Behavioral Results

- Seq. learning in **left** and **right**
  - Loc. right fastest
  - Loc. bottom slowest



## Key Question:

- Did our experimental paradigm show predictive learning behaviorally and without the confound of motor learning component?

## Summary

## Key Findings:

- **General learning;** Faster responses over time.
  - Fastest RTs in loc. right.
  - Slowest RTs in loc. top.
  - **Sequence-specific learning;** Quadratic trend over transition blocks in loc. **left**. and **right**. item

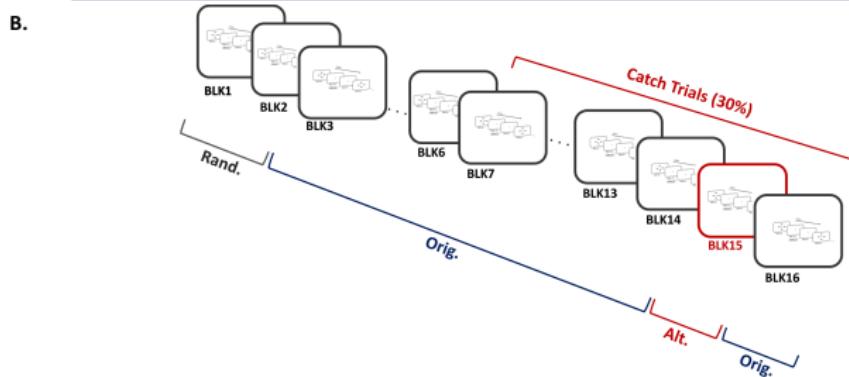
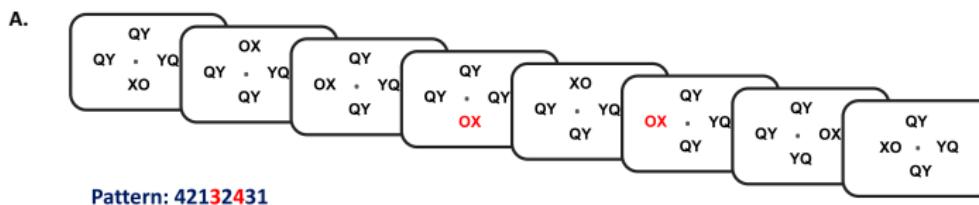
## Experiment 2



## Procedure

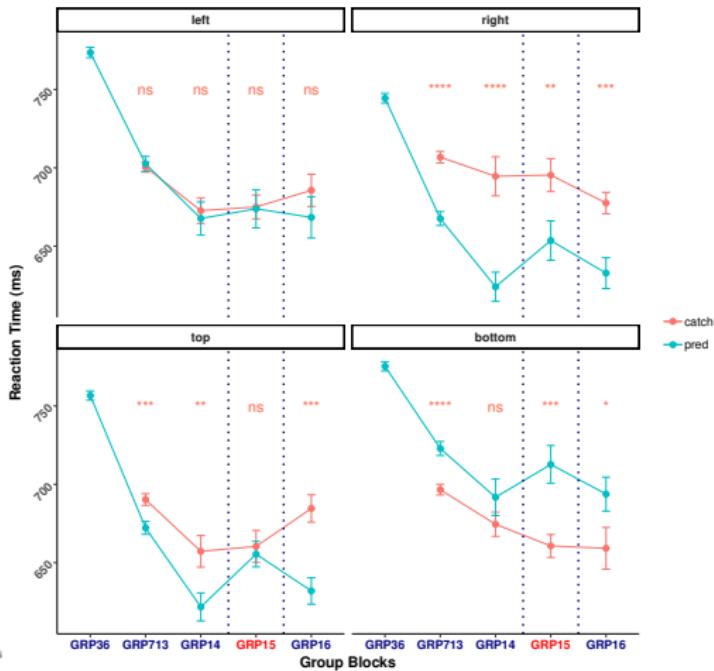
Goal

- Add **catch** trials to the experiment
    - Measuring a **trial-by-trial** predictive learning
    - Measure **prediction error** after recording EEG



# Behavioral Results

- Seq. learning in loc. right
- Loc. right fastest
- Loc. bottom slowest



## Key Question:

- Predictive learning behaviorally and without the confound of motor learning component?

## Summary

## **Key Findings:**

- **General learning;** Faster responses over time.
  - Fastest RTs in loc. right.
  - Slowest RTs in loc. top.
  - **Sequence-specific learning;** Quadratic trend; only in location **right** and **not left**.

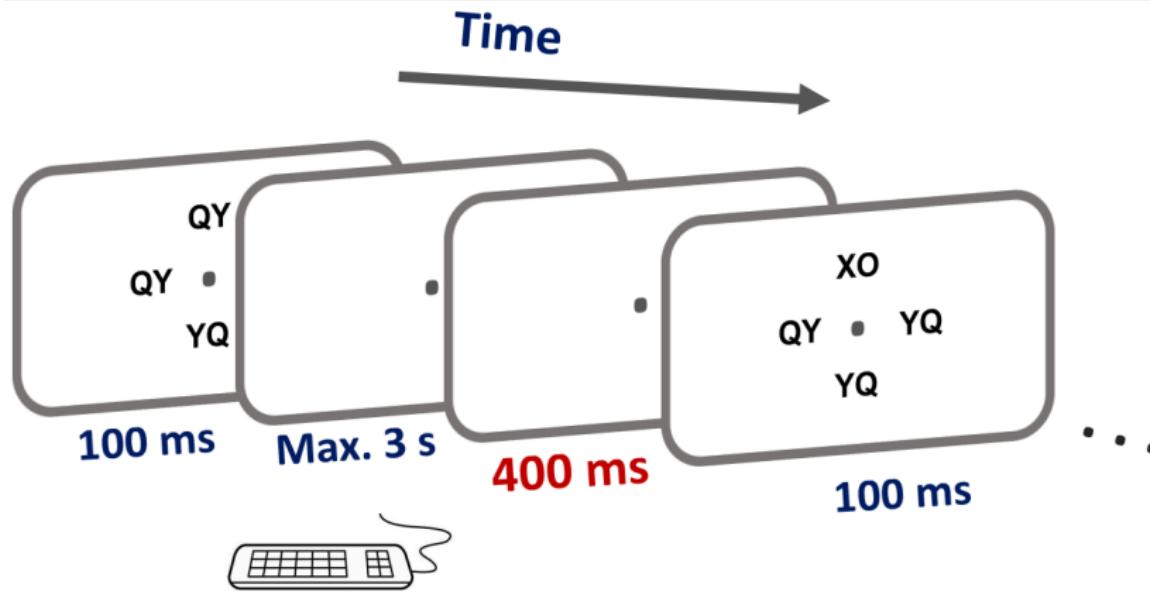
## Next step:

- After observing promising behavioral results, we recorded EEG.

## Summary

## Next step:

- Increasing baseline.

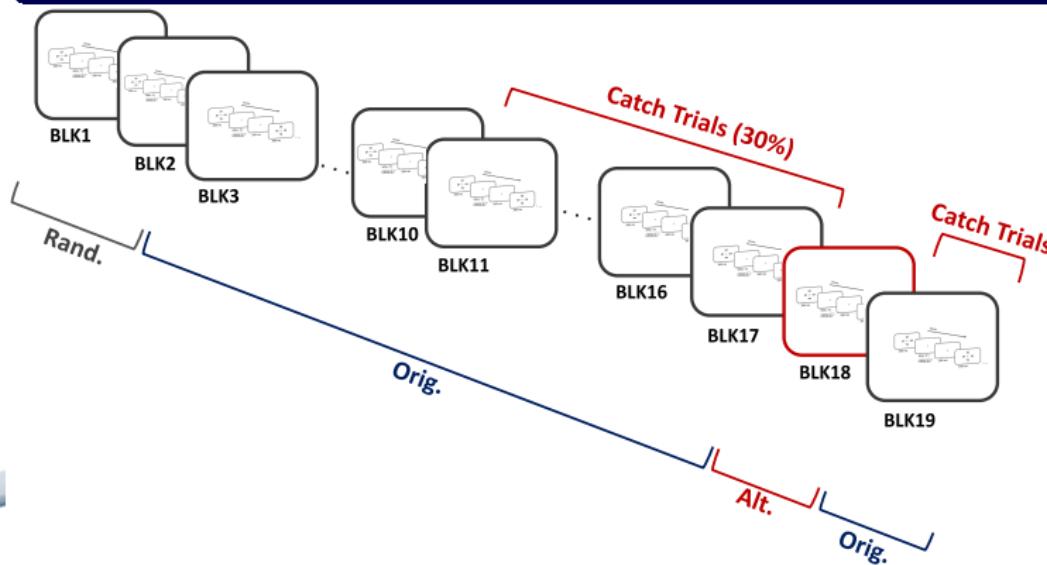


## Experiment 3

## Procedure

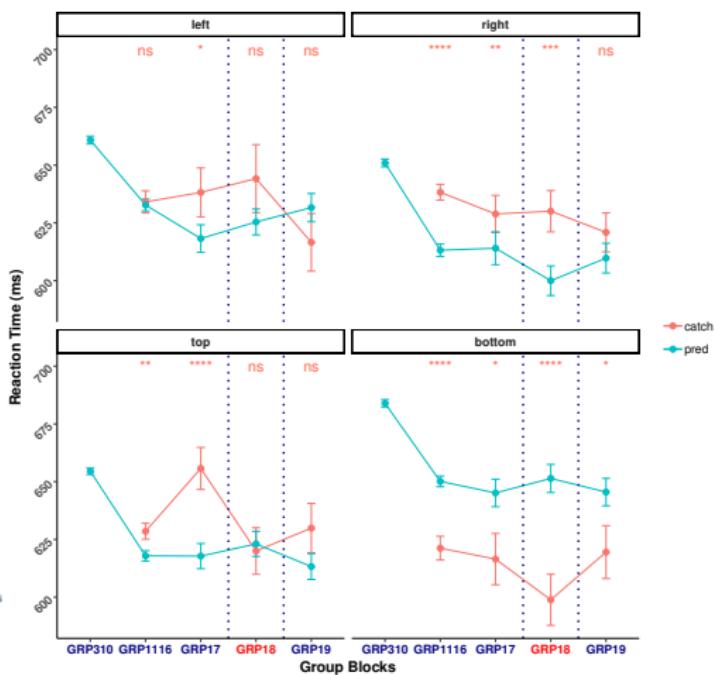
## Goal:

- Replicate the previous findings after increasing the ISI.



## Behavioral Results

- Seq. learning in loc. right
  - Loc. right fastest
  - Loc. bottom slowest



## Summary

### Key Findings:

- **General learning;** Faster responses over time.
- Fastest RTs in loc. right.
- Slowest RTs in loc. top.
- **Sequence-specific learning;** NO significant quadratic trend over transition blocks.

# EEG Analysis

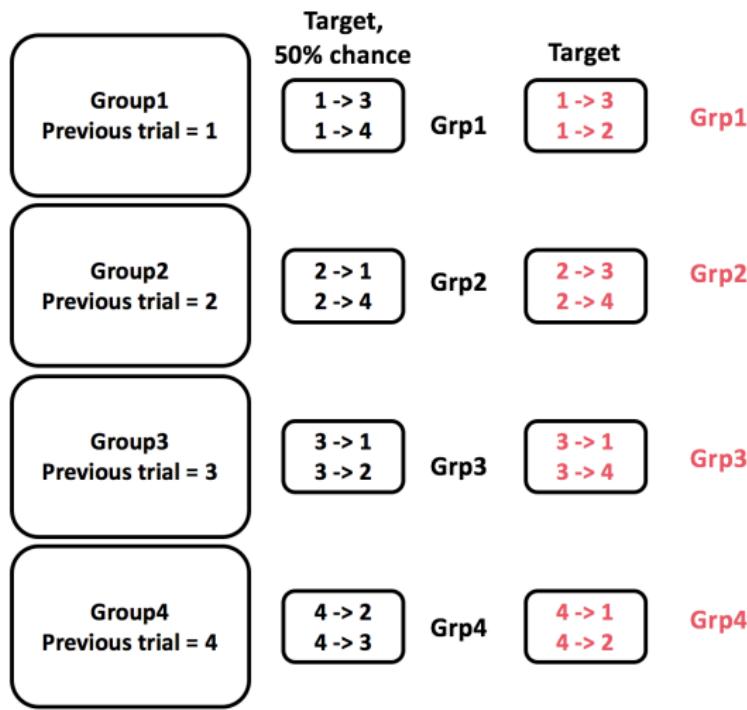
## Second-order feature

**2** **1** -> **3**

**3** **1** -> **4**

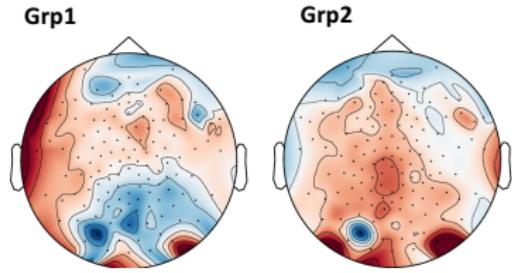
**Pattern S1:** 4213 2431

**Pattern S2:** 1342 3124

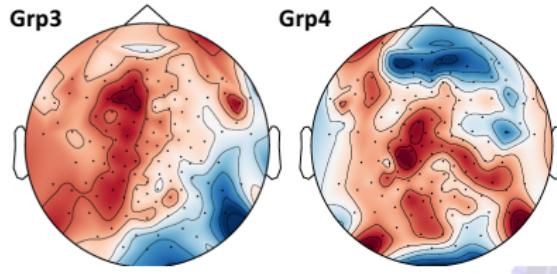
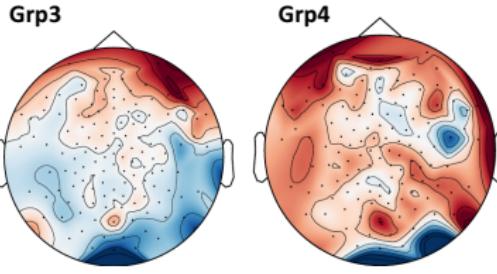
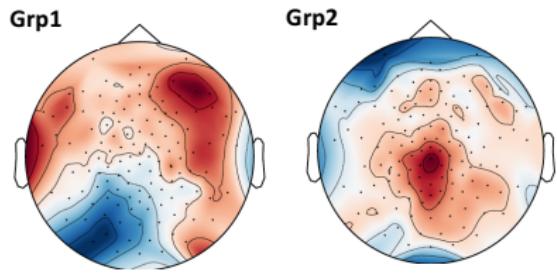


# Topographical maps

Pre-stimulus Block 3-6, P1

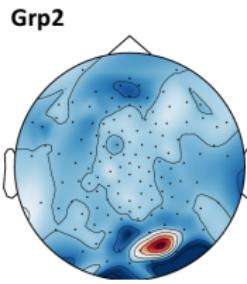
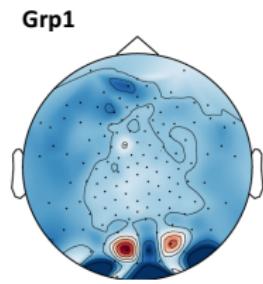


Post-stimulus Block 3-6, P1

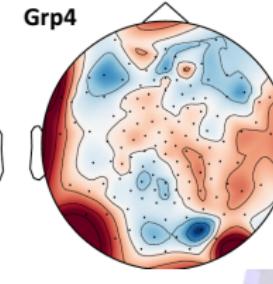
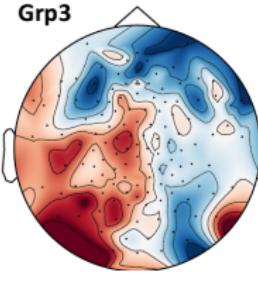
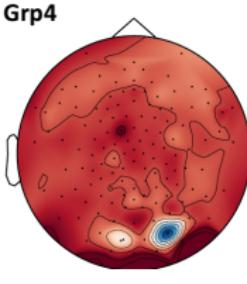
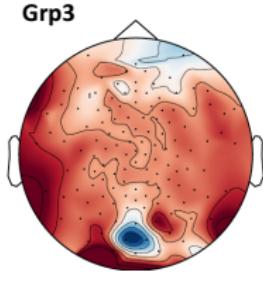
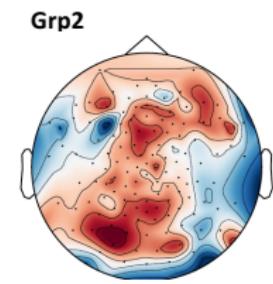
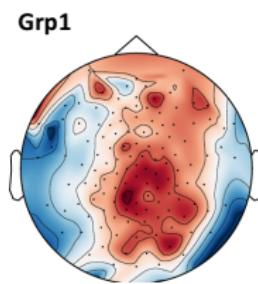


# Topographical maps

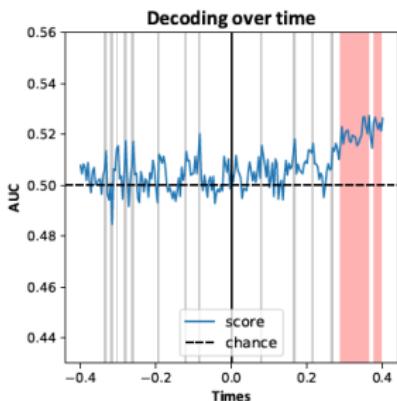
Pre-stimulus Block 3-6, P2



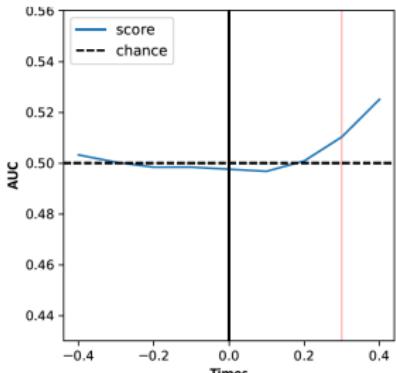
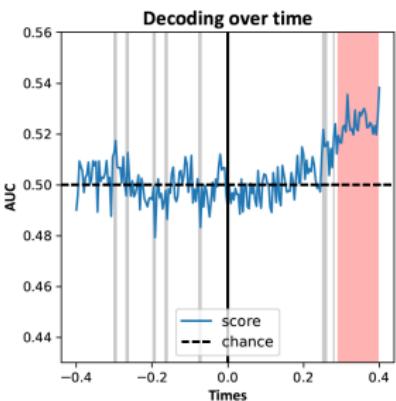
Post-stimulus Block 3-6, P2



## Block 3-6



## Block 7-10



# Temporal Generalization

- Decoding performance as a function of time that specifies when a certain piece of information becomes explicitly encoded in brain activity.
- Blind to dynamics of brain activity
- The next question we may ask is: **how does the underlying neural code evolve in time?**
- Generalization across time reveals how mental representations are dynamically transformed

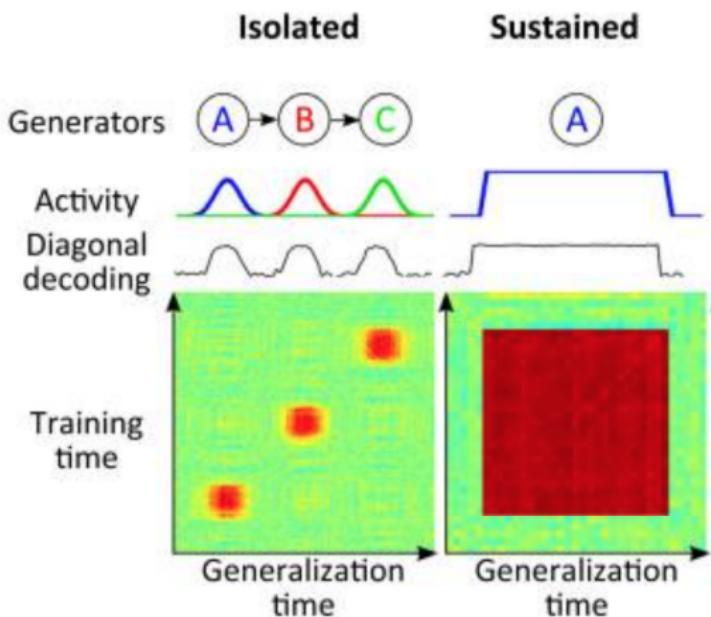
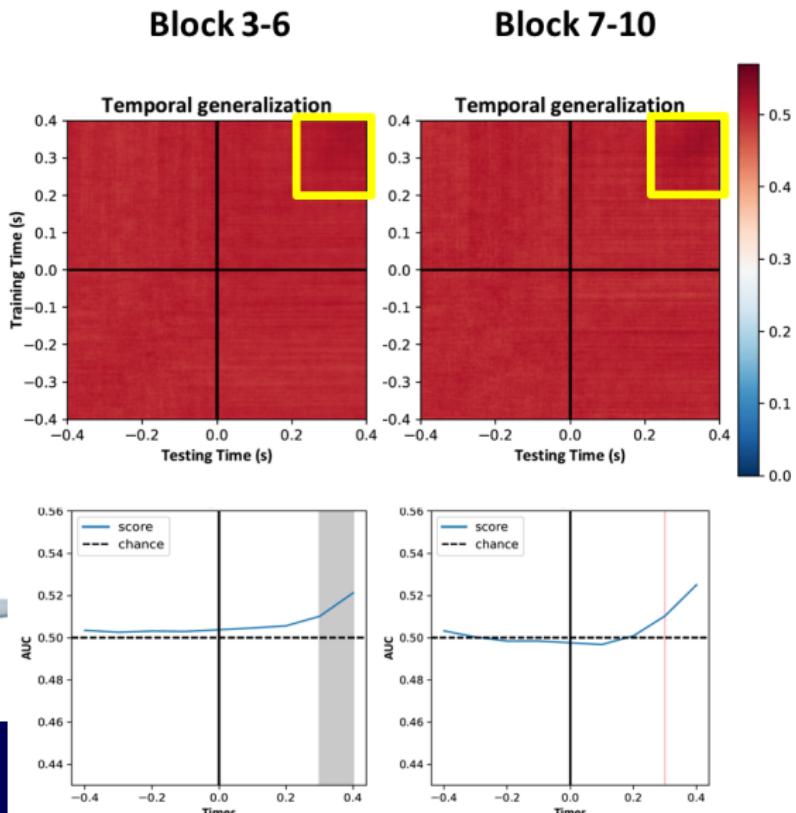
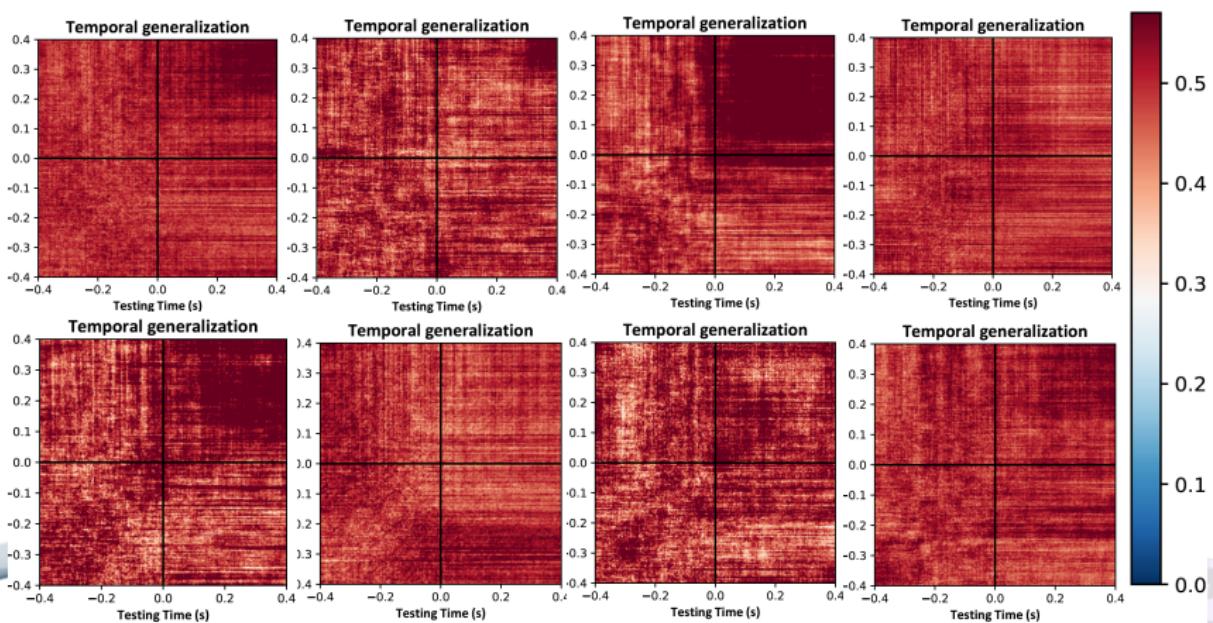


Figure: King, J.-R., Dehaene, S. Characterizing the dynamics of mental representations: the temporal generalization method. Trends Cogn. Sci. 18, 203–210 (2014).

# With Cross Validation

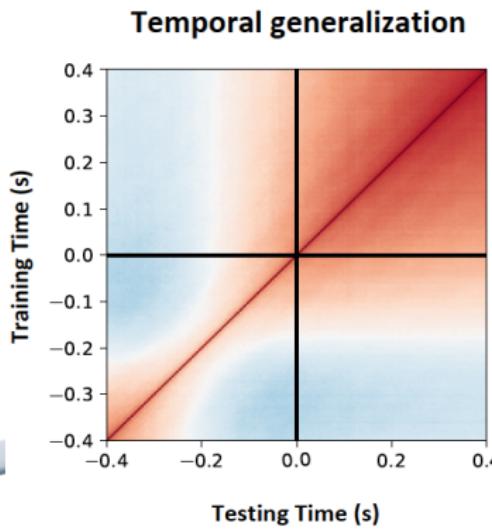


# With Cross Validation

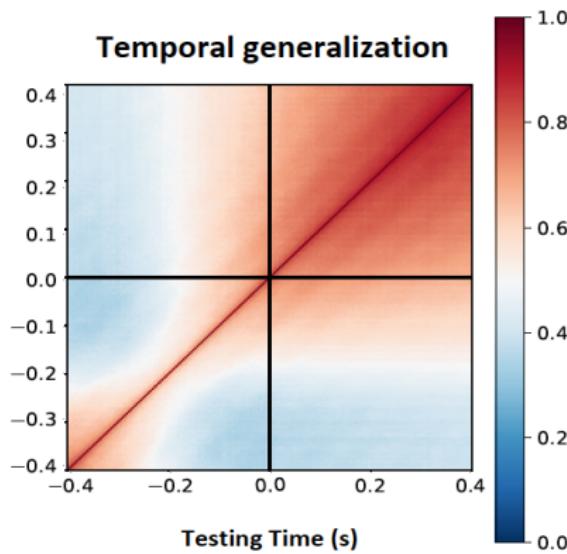


# No Cross Validation

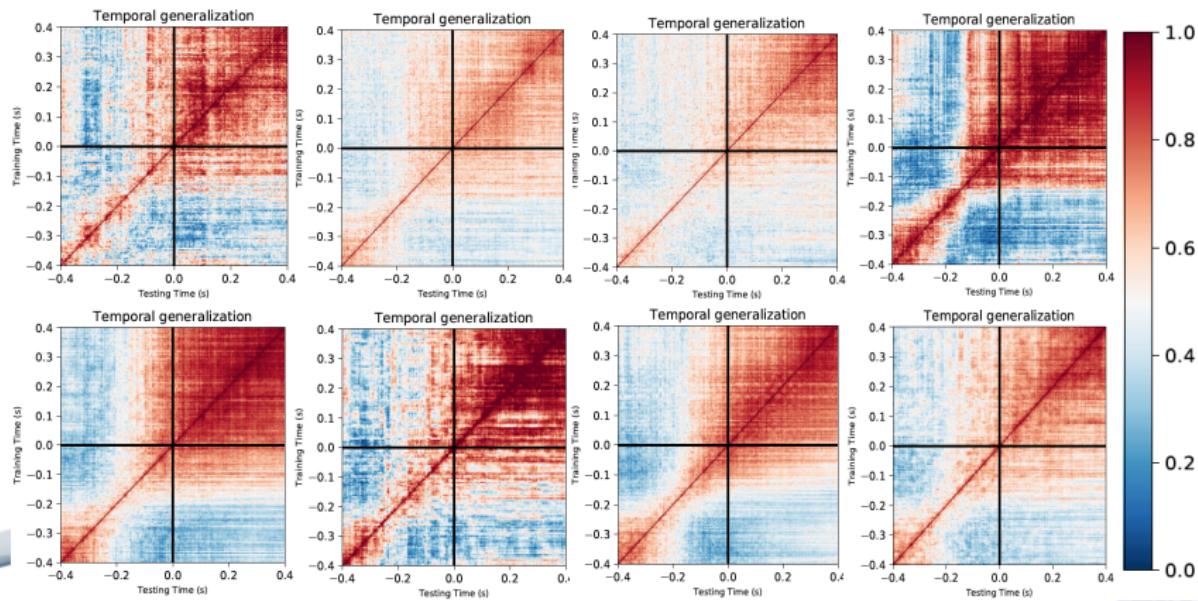
**Block 3-6**



**Block 7-10**



# No Cross Validation



## Summary

## Summary

### Key Findings:

- **General learning;** Faster responses over time.
- Consistent behavioral results over 3 experiments.
- Effect of ISI on pure perceptual learning
- EEG results showed a promising pattern of temporal generalization

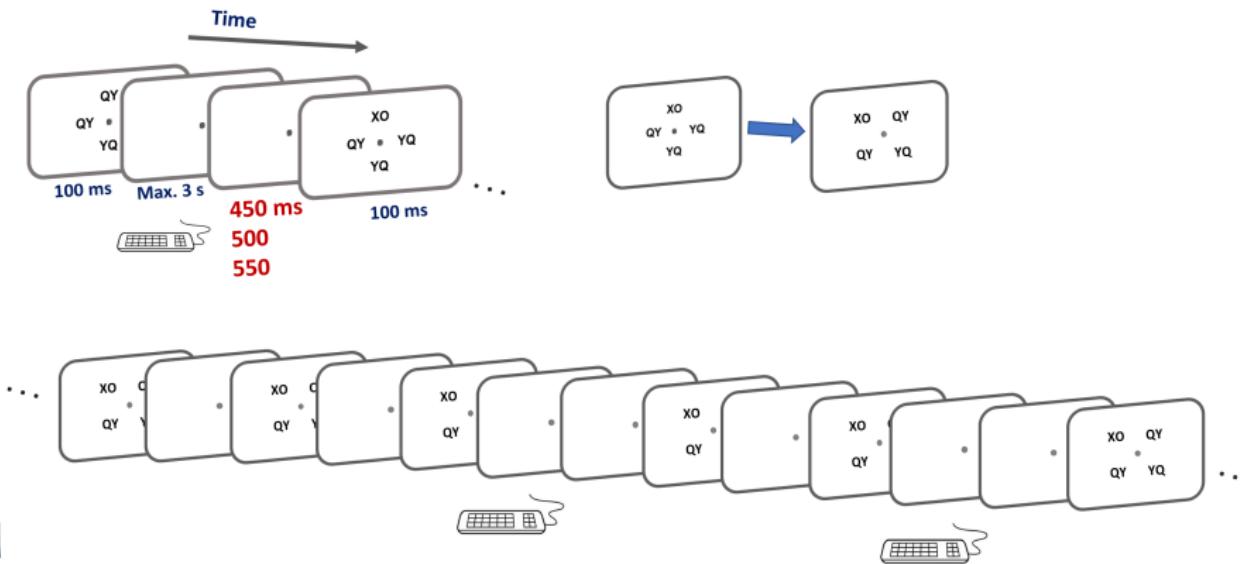
# Summary

## Challenges:

- Implicit learning
- Pure perceptual learning
- Longer ISI

## Next Experimental Paradigm

# New Experimental Paradigm



## Acknowledgement

## **My Supervisors**

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  - Ananta Nair
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  - Andrew Carlson
  - Alan Zheng



End!