

Where is Distraction?

The effect of bottom-up distraction in category learning

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1. Introduction

1-1. How people learn categorization?

- **Selective attention** : cognitive process of focusing on specific information while simultaneously suppressing irrelevant or distracting information.
- It operates **attending selectively** on **category-relevant information**, and **filtering out category-irrelevant dimensions**. (Reed 1972; Mendin & Shaffer 1978; Nosofsky 1986; Carroll & Chang, 1970)
- The representative cognitive model **ALCOVE** uses attention parameters($\lambda\alpha$) within a neural network framework to simulate this learning process. (Kruschke, 1992)

1-2. Developmental Differences

- A **cost of selective attention** occurs when previously irrelevant information later becomes relevant. (Best, Yim, & Sloutsky, 2013)
- This cost appears clearly in adults but not in children, which is called **diffused attention**. (Deng & Sloutsky, 2013)

2. Experiment

2-1. Participants

91 Participants (49 Females, M 21.81 yrs, SD 2.52 yrs)

2-2. Procedure

(A) Category Task

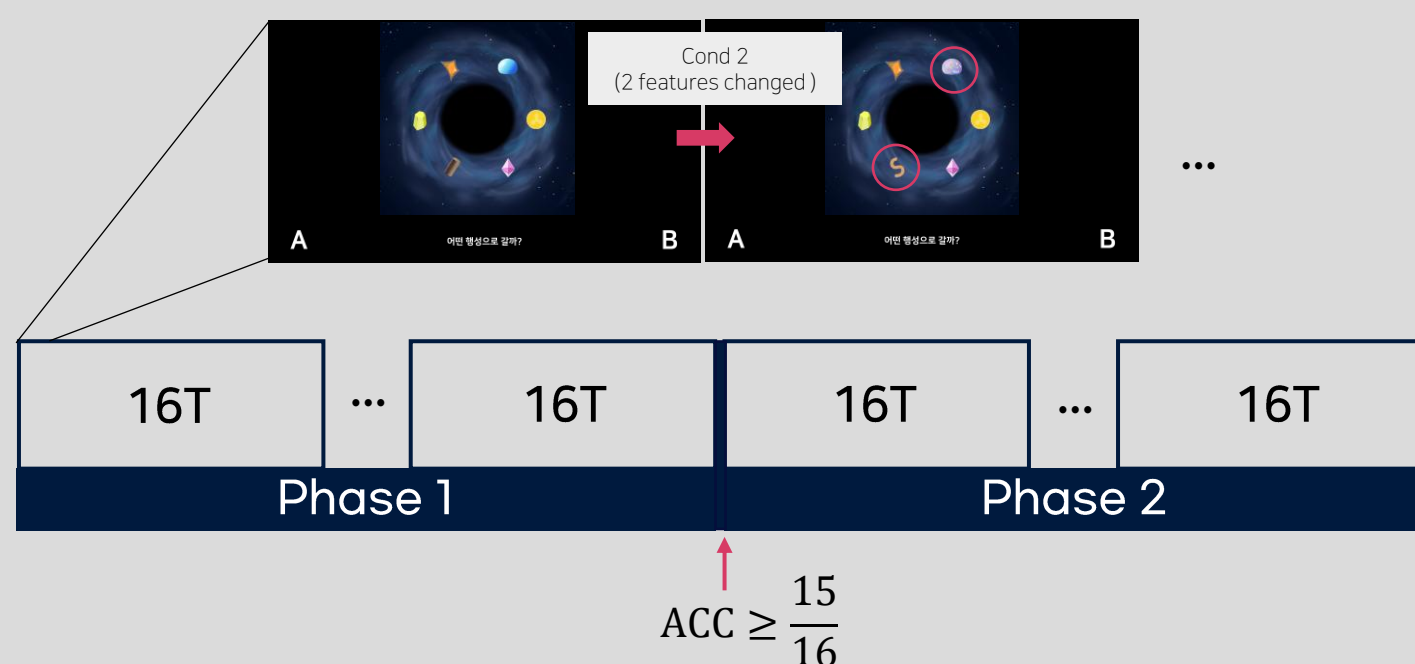
- Classify whether the wormhole stimuli belongs to planet A or B
- Participants receive feedback in each trials to find the categorization rule
- Eye gaze recorded with eye tracker (Eyelink 1000 plus), sampling rate 2K Hz

◆ Materials

- Wormhole Stimuli with 6 dimensions, each dimension is binary
- One is category-relevant dimension
- Condition (2 or 3): Number of features changed each trial (distraction \downarrow)

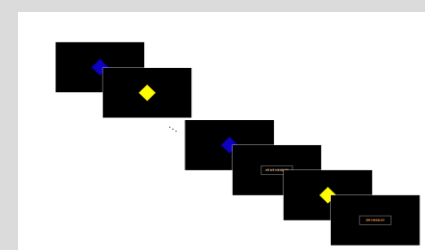
◆ Design

- 16 Trials in each block
- Category-relevant dimension change unbeknownst to the participants if 14 corrects per block
- Max 15 blocks participants can try

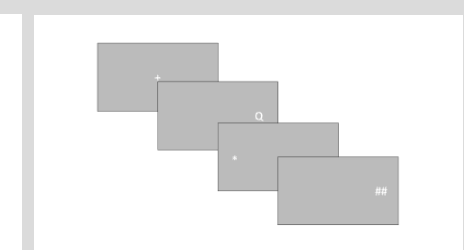


(B) Attentional Control Task

(a) Go/No-go Task



(c) Antisaccade Task



(b) Spacing flanker Task



(d) Symmetry Span Task



3. Conclusion

- Similar overall learning patterns of category learning task, differences in **selective attention cost** ($p = .098$), **distraction** ($p = .041$), and **entropy** ($p = .084$).
- A **significant correlation** with the attentional control task (Antisaccade) only in the Condition 3 ($p = .002$). → Attentional control ability becomes critical when the task is more demanding (not under easier tasks).
- Even with identical stimuli and the same information, learning performance can vary depending on the level of distraction. Faster learning occurs when the learning sequence remains stable.

► Contribution

- **Experiment methodological implication**
Proposes a novel experimental design, which can measure and dissociate distraction factor.
- **The need to extend existing cognitive models**
A new model that incorporates filtering (ignoring distraction) mechanisms is needed to more accurately account for human category learning.
- **Applicability in practical educational environment**
Design effective learning sequences for those who have low or impaired attention function (infants, the aged, ADHD, etc.). Guide more targeted instructional strategies.

► Where is Distraction?

Previous models primarily explain how attention is allocated, but they do not adequately address **filtering(distracton)**, which may be critical in real category learning situations.

► Research Goal

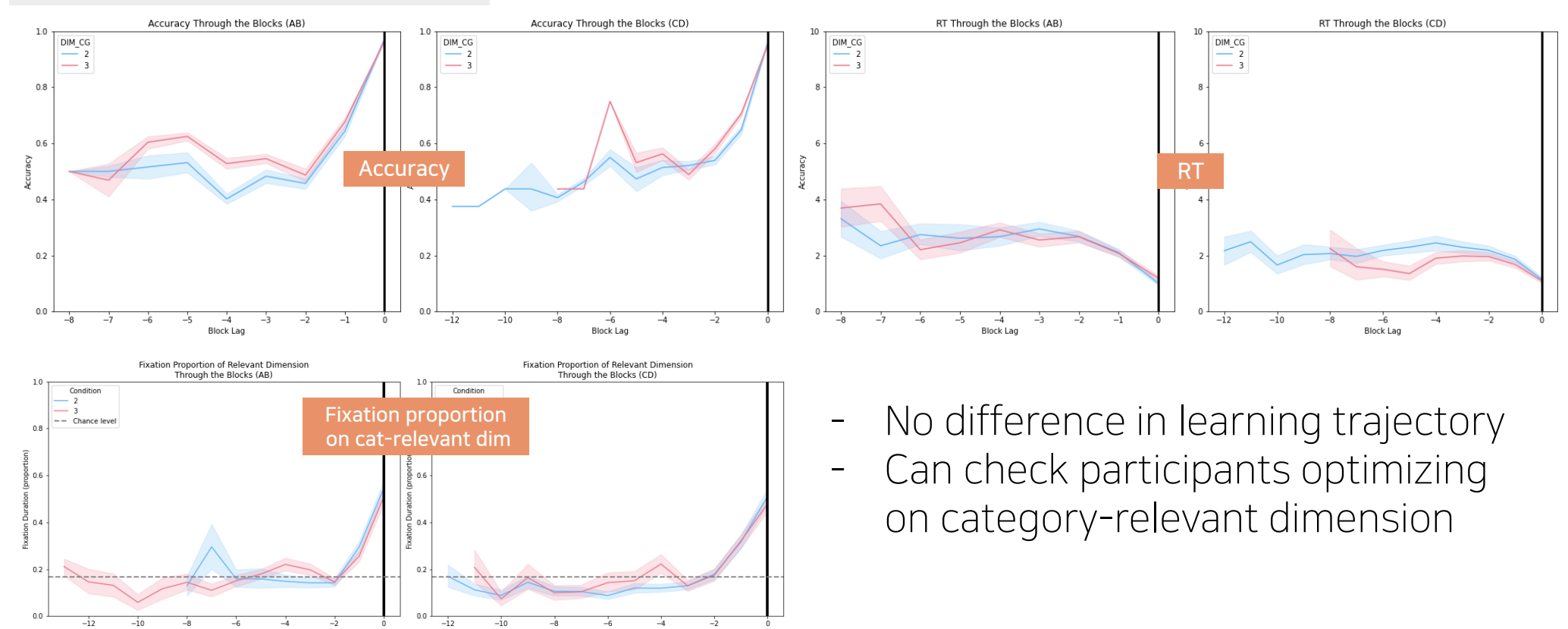
- Develop an experimental design that dissociates filtering in adult category learning
- Examine individual differences in filtering ability via correlations with an attentional control task
- Evaluate the need to extend existing cognitive models of selective attention.

3. Results

3-1. Learning

- **Learners**
Cond 2 (7 out 46 fail), Cond 3 (12 out 50 fail)
- **Number of Trials consumed to learn in Phase 1**
Cond 2 ($M = 56$, $SD = 30.72$) < Cond 3 ($M = 68.32$, $SD = 49.41$)
 $p = .103$, Student's t -test
- **Cost of selective attention (Differenced in Trials consumed in each phase)**
Cond 2 ($M = -17.231$, $SD = 60.53$) < Cond 3 ($M = 0.571$, $SD = 37.58$)
 $p = .098$, Student's t -test

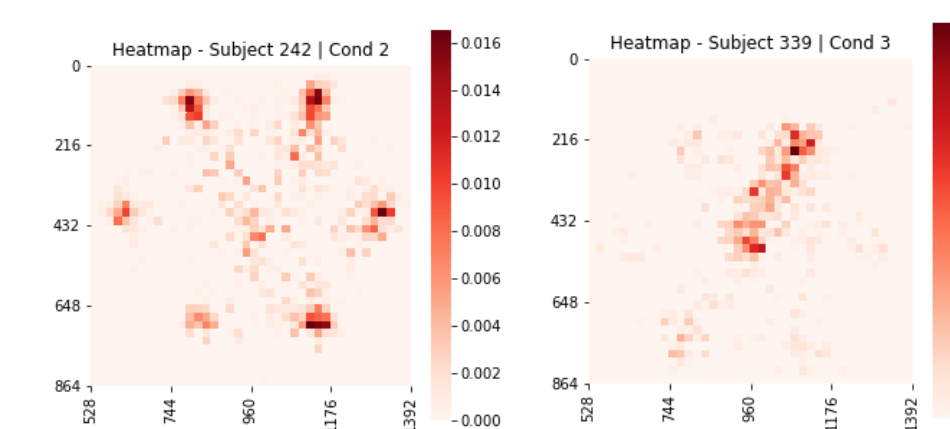
3-2. Optimization



- No difference in learning trajectory
- Can check participants optimizing on category-relevant dimension

3-3. Distraction

- **Fixation Count Heatmap**
Difference in distraction Cond 2 vs 3.
- **Distraction / Entropy**
Cond 2 < Cond 3

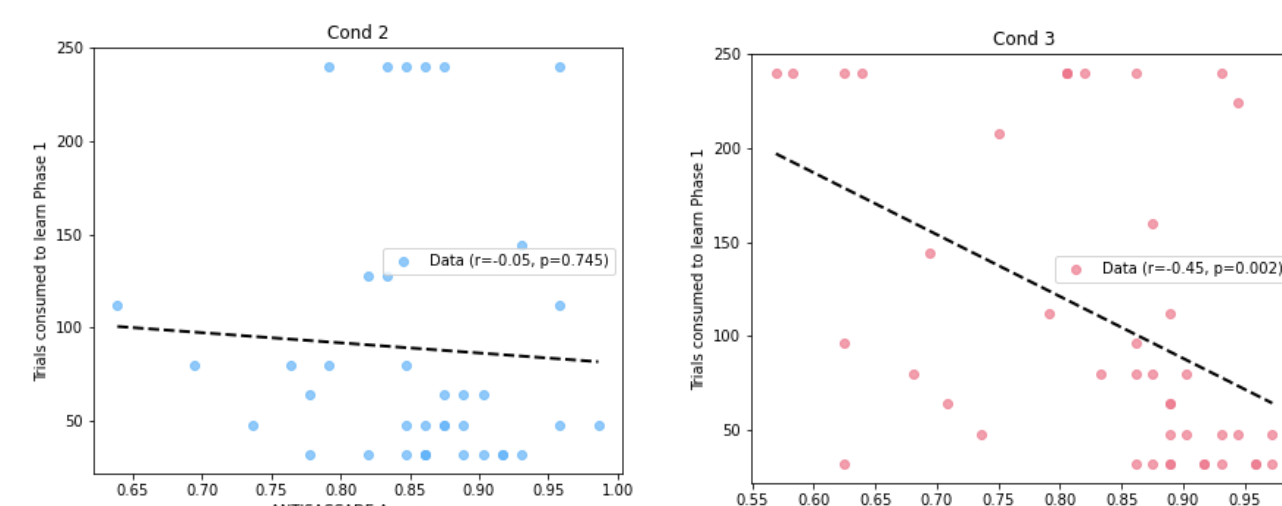


	t	p
DISTRACT	1.775	.041
Entropy	1.396	.084

*Distraction: Fixation duration on looking changed dimensions on first 2 block
*Entropy: Entropy of fixation

3-4. Attention Correlation

- Antisaccade task & Number of trials consumed to learn Phase 1



Only Cond 3 showed correlation.
($p = .002$)

3-5. ALCOVE Model Simulation

- Parameter: $c = 5$, $\phi = 2$, $\lambda\alpha = 1.0E-2$, and $\lambda\omega = 5.0E-2$
- No difference between Cond 2 vs 3

