

How surprising information shifts our attention and influences learning

Em Heffernan

Category learning: grouping items by attending to shared features

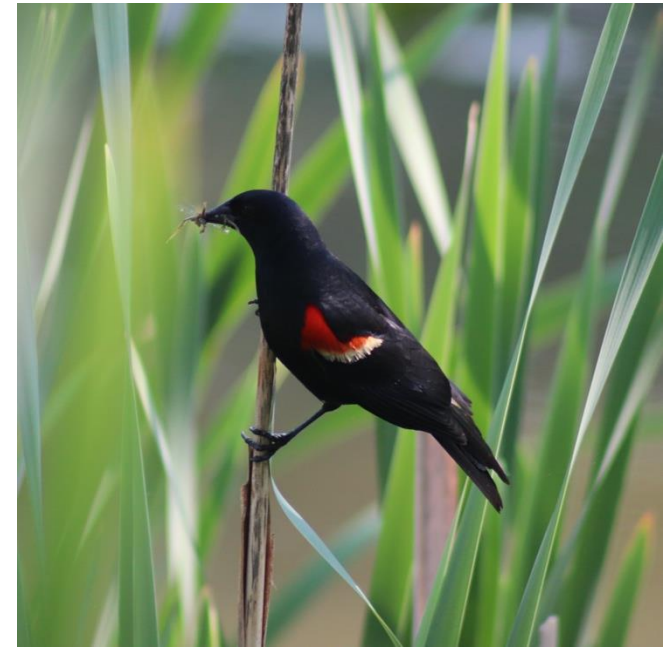
Category learning is an imperfect process. Because they look similar, we might assume that the center bird is related to the bird on the left, but it's actually a blackbird!

savannah sparrow

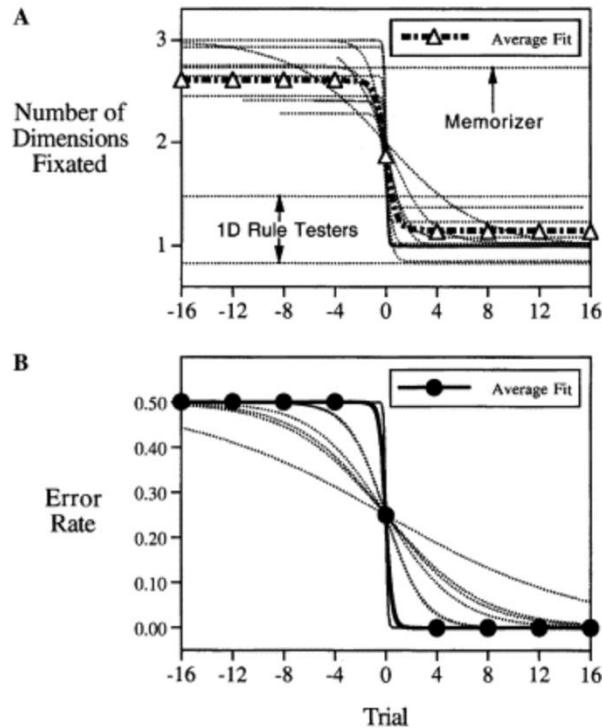


another sparrow??
female red-winged black bird

red-winged black bird



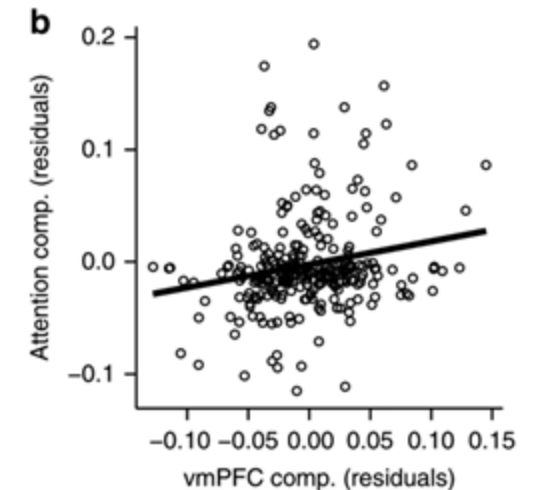
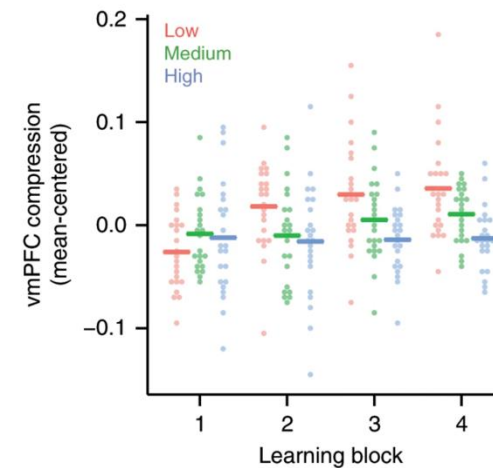
Attentional tuning in category learning



(Rehder & Hoffman, 2005)

We can measure attention during learning via eye or mouse-tracking

(Rehder & Hoffman, 2005; Blair et al., 2009; Zaki & Salmi, 2009; *Chen et al., 2013...)



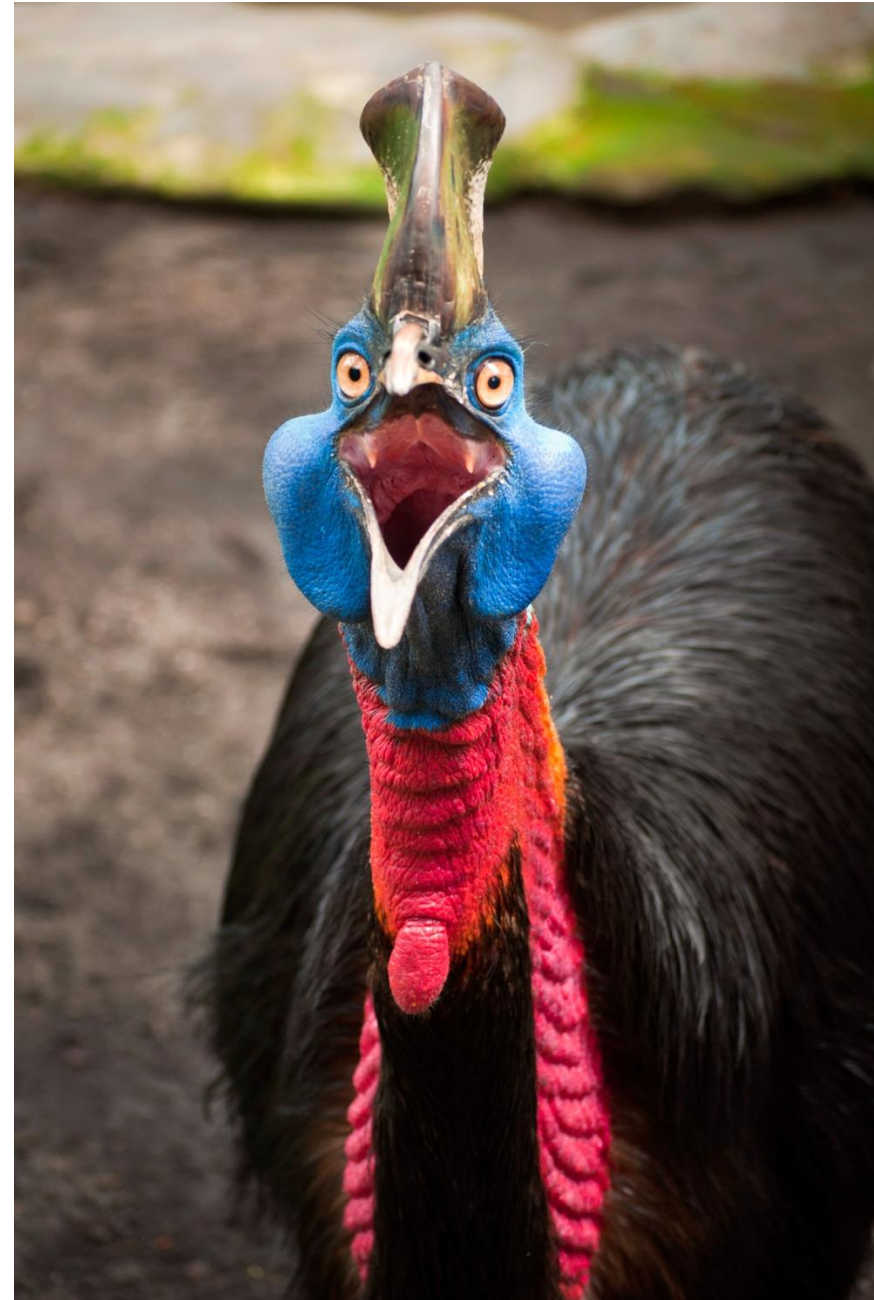
We can also measure attentional compression in the brain

(Mack, Preston, & Love, 2020)



The ability to rapidly reallocate attention should be associated with better learning outcomes. However....

Not all exceptions
are created equal!



Cassowary (Ghetty Images)

Crossover exceptions



- Similar to **existing knowledge**, confusable
- Elicit prediction error, cause uncertainty
- Must be distinguished from rule-followers

Oddball exceptions



- Unique features, distinct from **existing knowledge**
- Novelty without prediction error
- Stored in isolation

The ability to rapidly reallocate attention should be associated with better learning outcomes
when exceptions overlap with existing knowledge

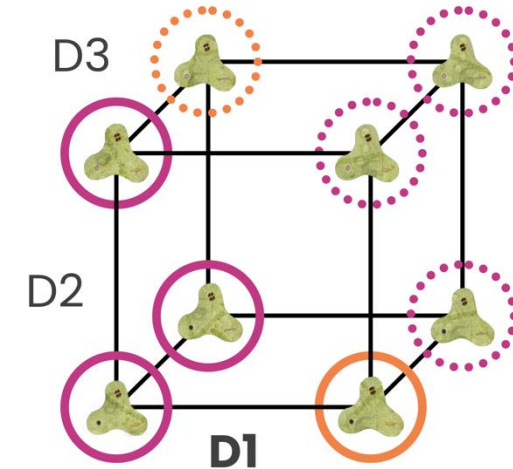
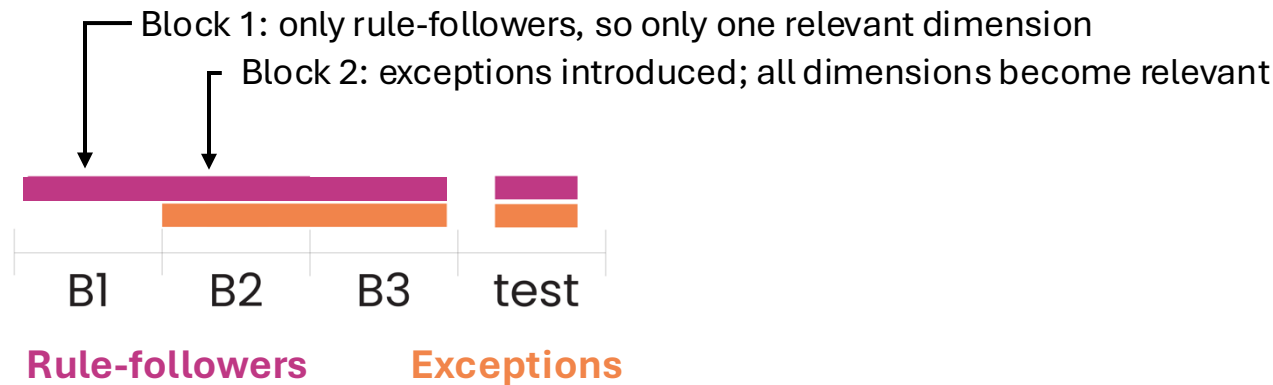
Experiment design



Xuan (Sophia) Zhang

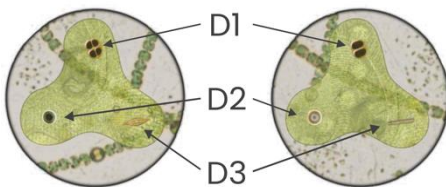


Marian Wang



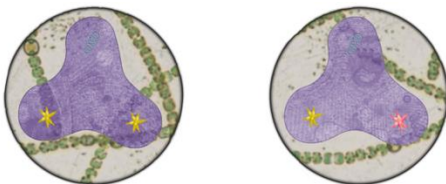
beta gamma

(Shepard, Hovland, & Jenkins, 1961 Type IV)



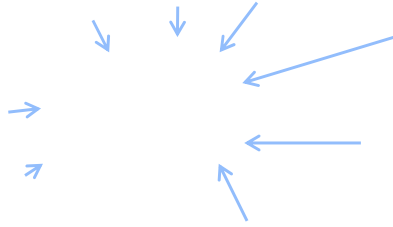
(Blair et al., 2009)

Crossover exceptions: Similar to existing knowledge, confusable
Elicit surprise, prediction error, & uncertainty

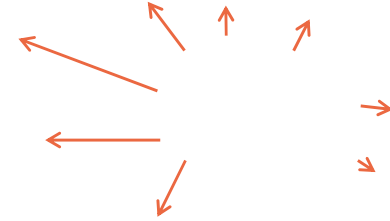
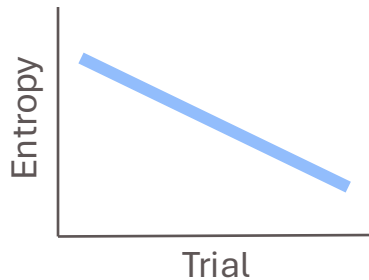


Oddball exceptions: Distinct from existing knowledge
Novelty without prediction error

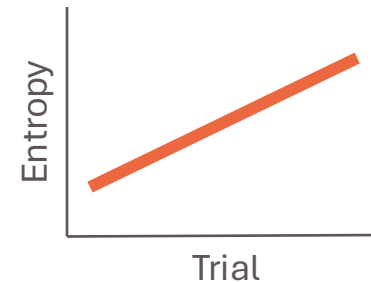
In Information Theory, **entropy (H)** is the amount of information conveyed in an event (Shannon, 1948)



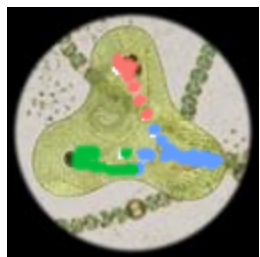
During rule learning, entropy (calculated from mouse-tracking data) should **decrease** as one fixates on fewer features



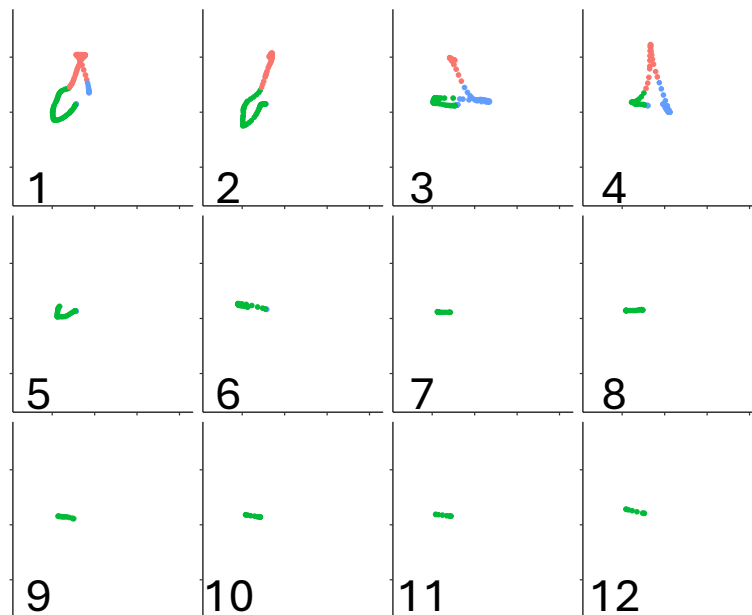
Following (crossover) exception introduction, entropy should **increase** as attention expands



We can use entropy to track attentional tuning before and after exception introduction



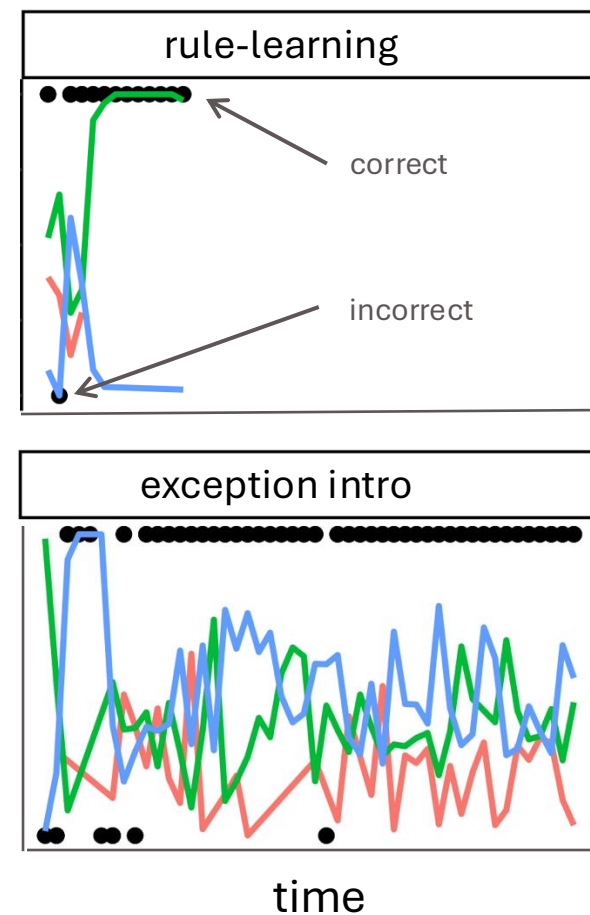
Mouse-tracking
data



Trial-by-trial
(rule learning)



fixation proportion



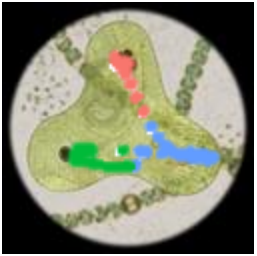
High entropy



Low entropy

Only confusable exceptions elicit widespread shifts in attention

Entropy as a measure of attentional distribution



High entropy Low entropy

Crossover introduction

LB1: Attention contraction

LB2: Widespread attention expansion

LB3: Attention optimization

Oddball introduction

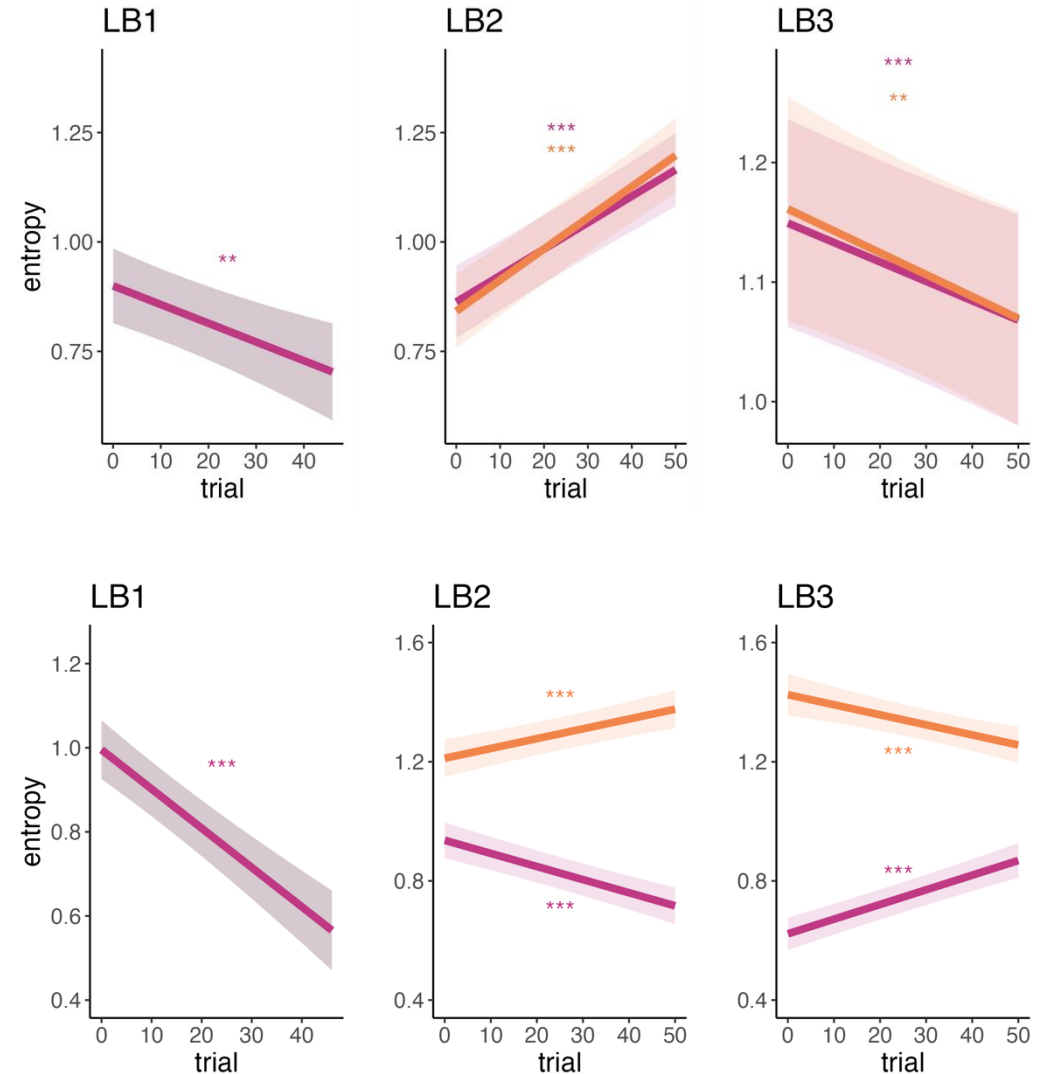
LB1: Attention contraction

LB2: Exception-specific expansion

LB3: Potential fatigue?

rule-followers

exceptions

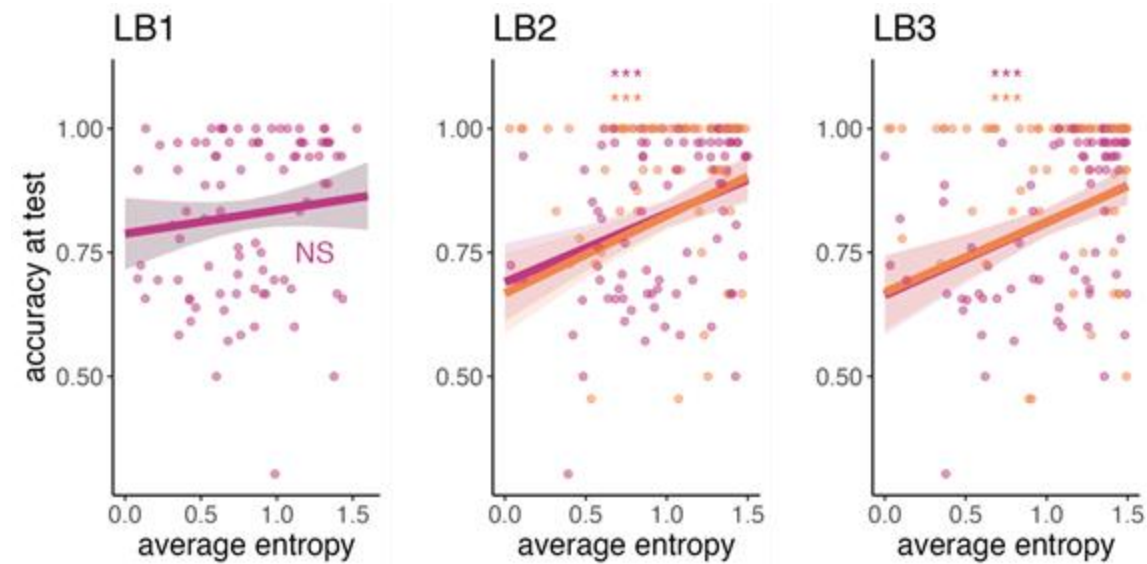


Learning confusable exceptions demands attention expansion

Crossover introduction

LB1: No relation

LB2&3: Increased entropy
associated with
increased accuracy

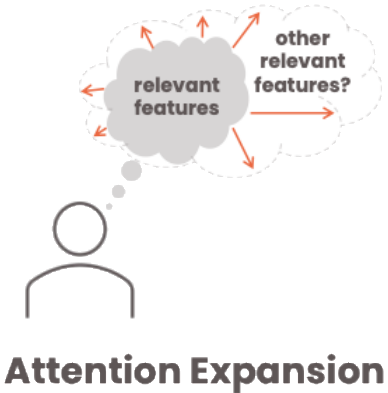
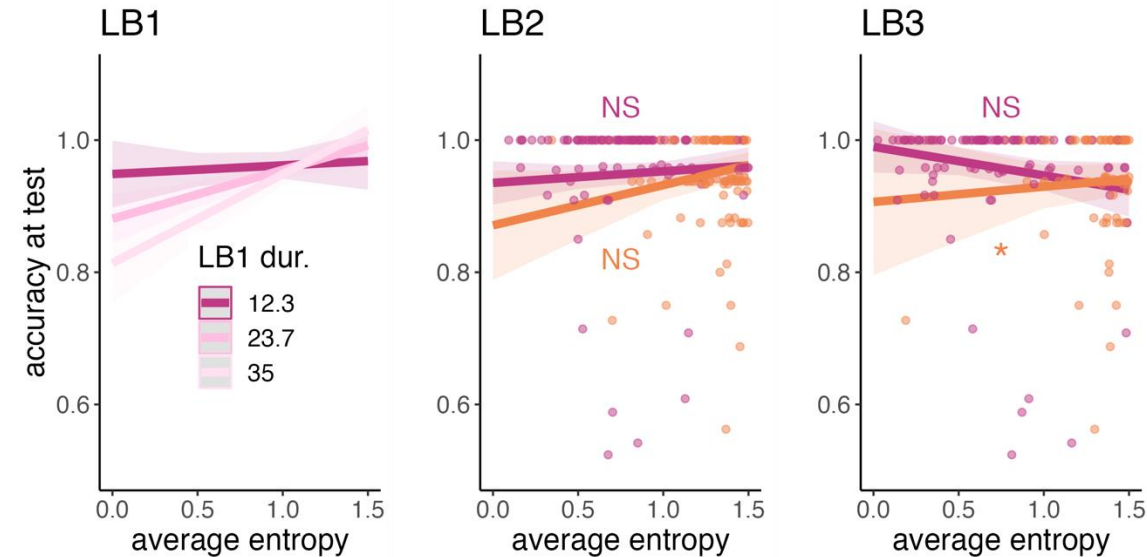


Oddball introduction

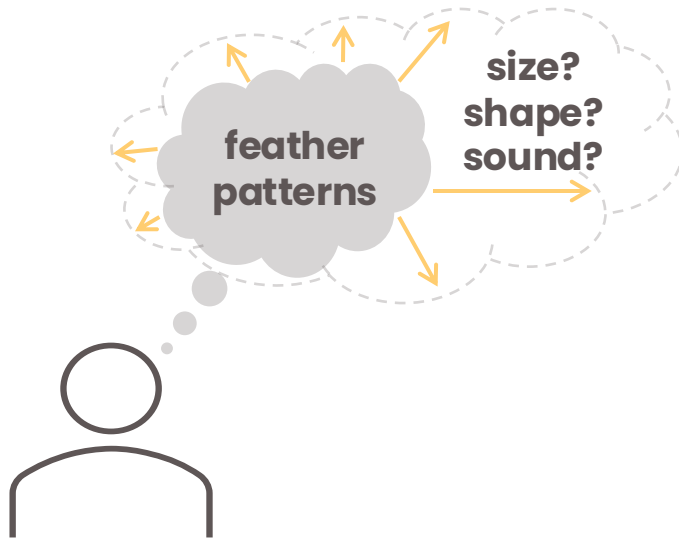
LB1: Duration-dependent
relation

LB2: No relation

LB3: Positive relation for
exceptions



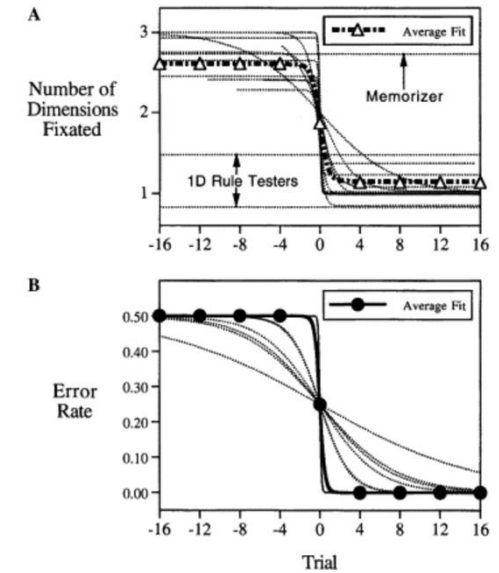
Some Takeaways



Crossover (but not oddball) exceptions lead to adaptive expansion of attention



How surprising information overlaps with existing knowledge is important for attention and learning



This work corroborates existing findings on how uncertainty mediates selective attention.

Thank you!



<http://macklab.utoronto.ca>

Michael Mack	Yongzhen (Dory) Xie
Frida Printzlau	Sanjivan Lognathan
Sagana Vijayarajah	Selina Fu
Gaeun Son	Rafidal Islam
Mateja Perovic	Marian Wang
Melisa Gumus	Xuan (Sophia) Zhang

And members of Budding Minds Lab

