

# Conservativity and Learnability of Determiners

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TIM HUNTER  
*Yale University*

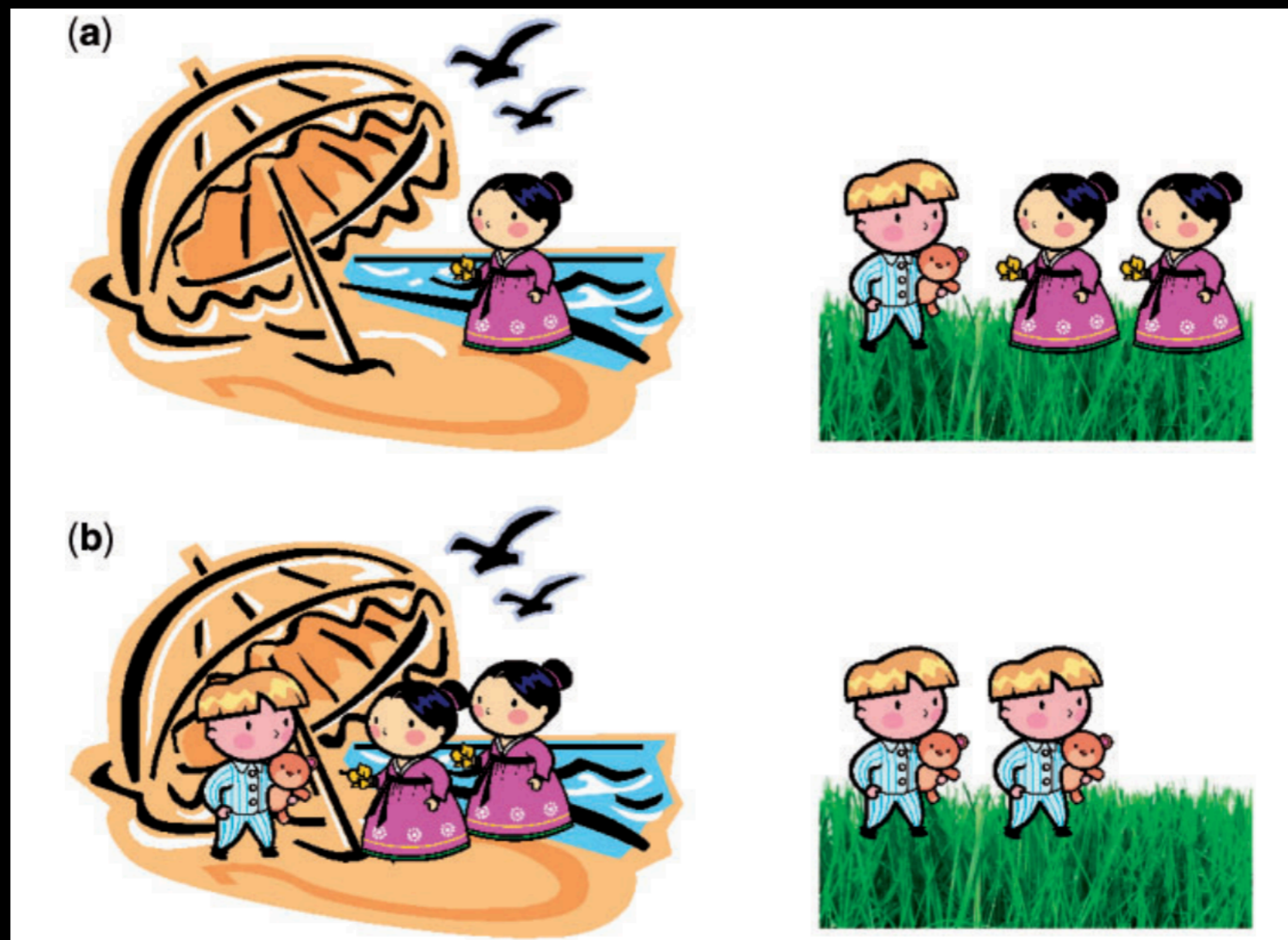
JEFFREY LIDZ  
*University of Maryland*

## **Abstract**

A striking cross-linguistic generalisation about the semantics of determiners is that they never express non-conservative relations. To account for this one might hypothesise that the mechanisms underlying human language acquisition are unsuited to non-conservative determiner meanings. We present experimental evidence that 4- and 5-year-olds fail to learn a novel non-conservative determiner but succeed in learning a comparable conservative determiner, consistent with the learnability hypothesis.

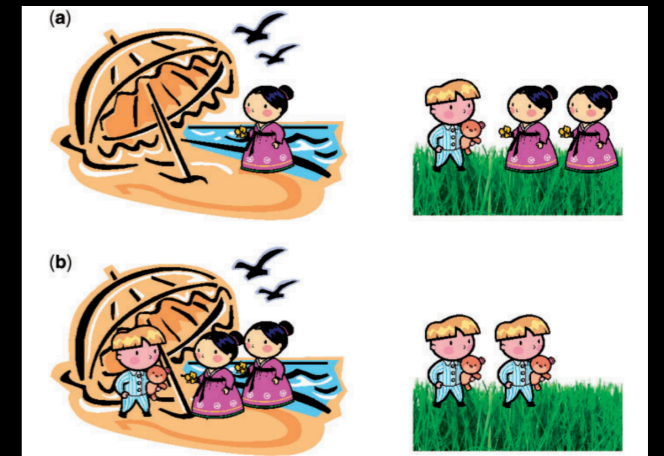
# Gleeb vs Gleeb'

- Gleeb = 'not all'
- Gleeb' = 'not only'



# Lidz & Hunter's experiment

- Two conditions: CONS and non-CONS
- Picky puppet task (Waxman & Gelman 1986).
- Warm-up (3 cards) - Training (5 cards) - Target (5 cards)
  1. *The puppet told me that he likes this card because gleeb girls are on the beach*
  2. *The puppet told me that he doesn't like this card because it not true that gleeb girls are on the beach.*

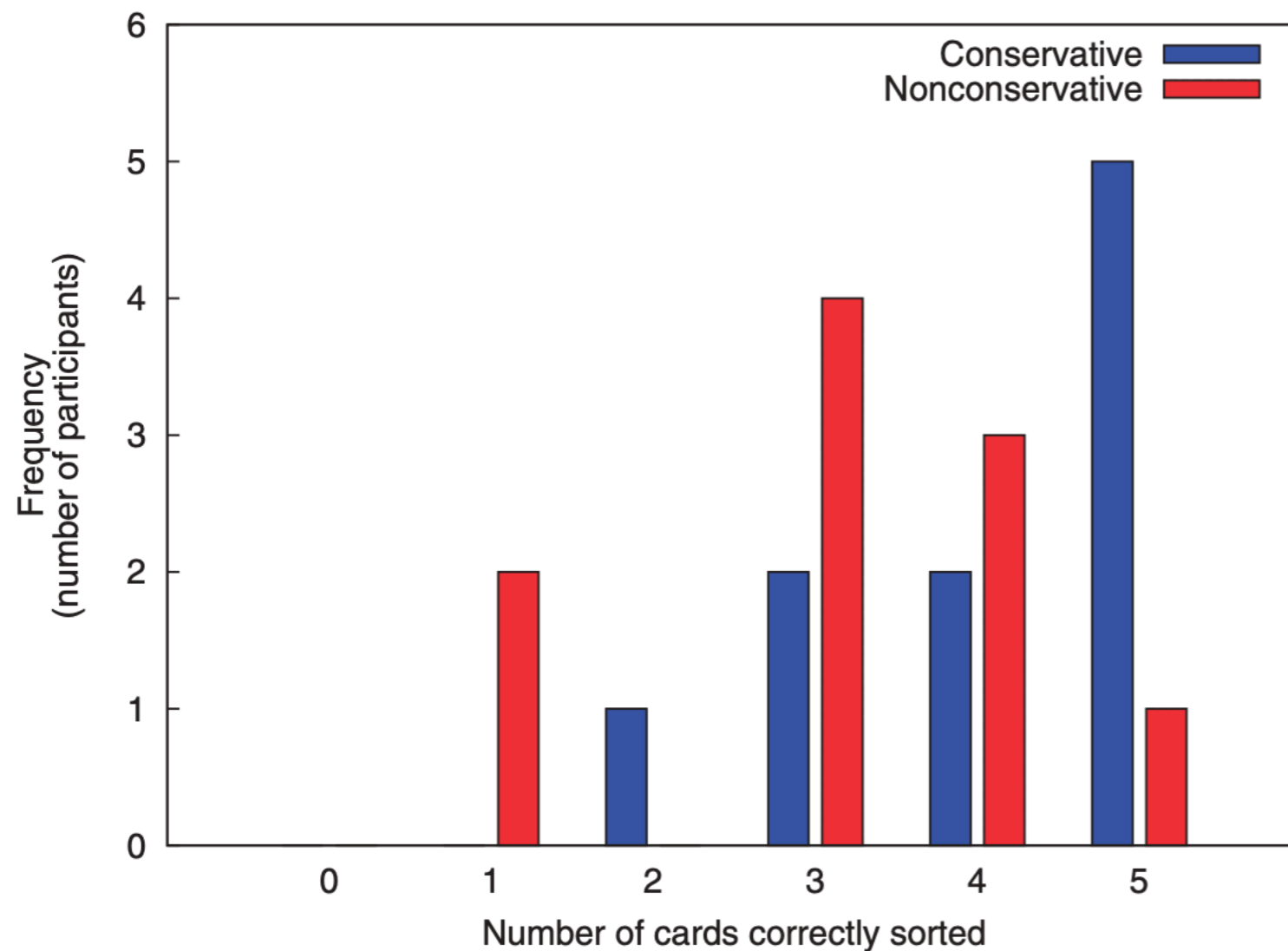


# Participants

- 20 children
- Aged 4.5 to 5.6 (mean 5.0)
- Conservative condition 4.5 to 5.5 (mean 4.11)
- Non-conservative condition 4.11 to 5.3 (mean 5.1)

# Results

Condition	Conservative	Non-conservative
Cards correctly sorted (out of 5)	mean 4.1 (above chance, $p < 0.0001$ )	mean 3.1 (not above chance, $p > 0.2488$ )
Subjects with “perfect” accuracy	50%	10%



**“the puppet was confused about which characters  
on the cards were boys and which were girls”**

*-Perfect non-conservative kid; interpreting conservatively?*

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# Discussion

- Are kids interpreting *gleeb* as a GQ?
- Are the results consistent with a structural account?
- Do the *gleeb* and *gleeb'* sentences differ only in conservativity? (Cf minimal pair discussion)
- Unclear that it replicates (Spenander and de Villiers 2019)
- Also, what about the other universals?

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# Connecting Content and Logical Words

**Emmanuel Chemla and Brian Buccola**

Laboratoire de Sciences Cognitives et Psycholinguistique, Département d'Études Cognitives, ENS, EHESS, CNRS, PSL University, Paris, France

Email: chemla@ens.fr, brian.buccola@gmail.com

**Isabelle Dautriche**

Centre for Language Evolution, University of Edinburgh, Edinburgh, UK

Email: isabelle.dautriche@gmail.com

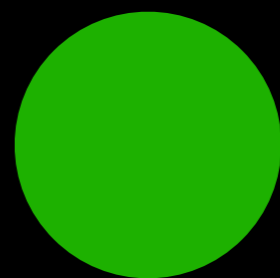
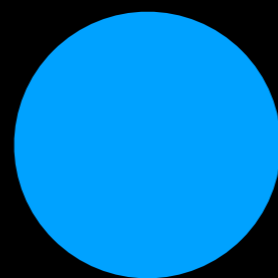
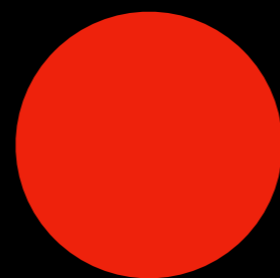
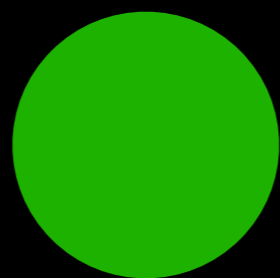
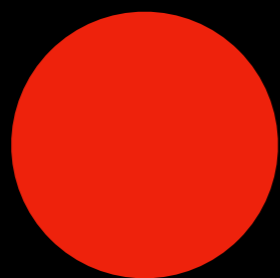
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## Abstract

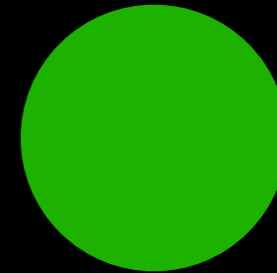
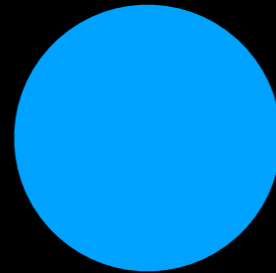
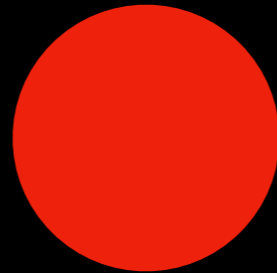
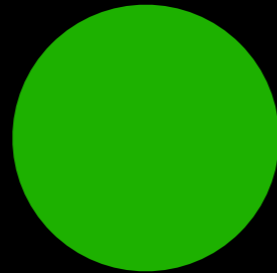
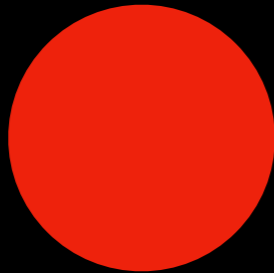
Content words (e.g. nouns and adjectives) are generally connected: there are no gaps in their denotations; no noun means 'table or shoe' or 'animal or house'. We explore a formulation of connectedness which is applicable to content and logical words alike, and which compares well with the classic notion of monotonicity for quantifiers. On

**Experiment Time!**

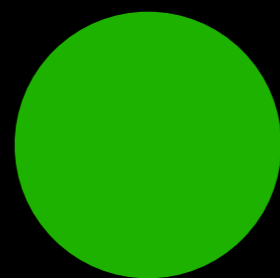
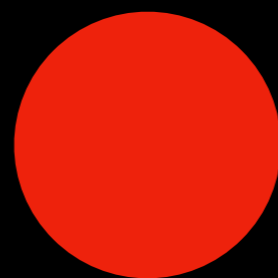
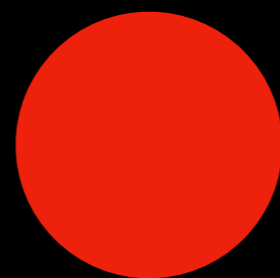
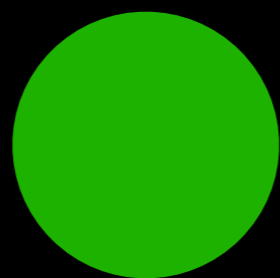


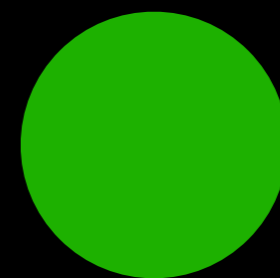
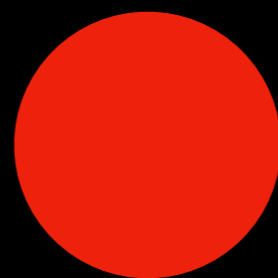
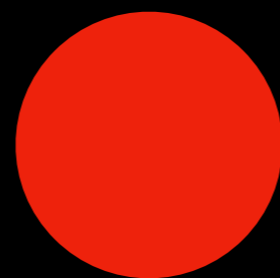
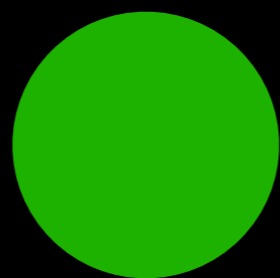
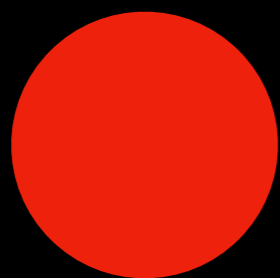




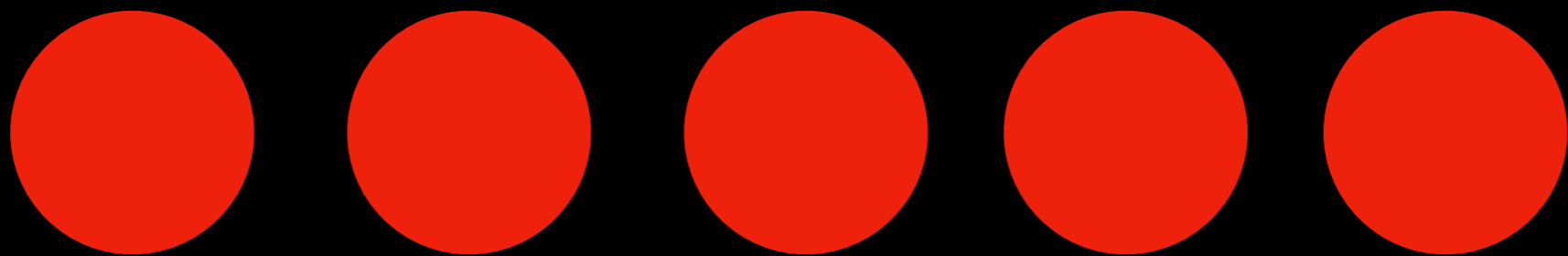


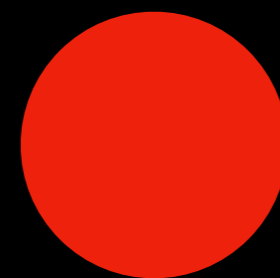
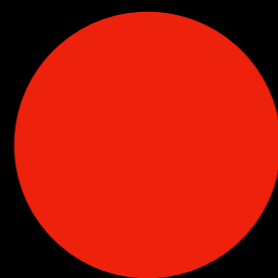
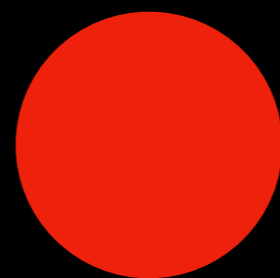
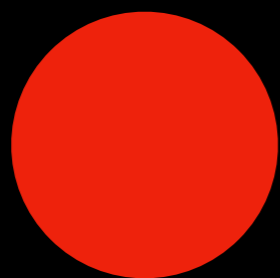






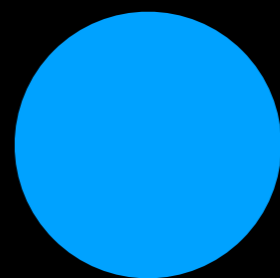
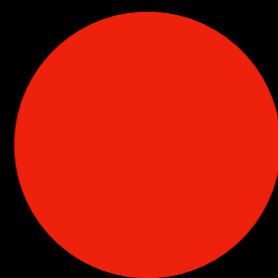
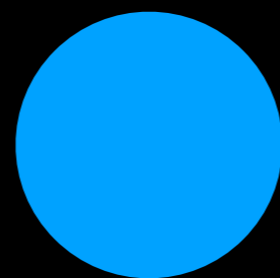
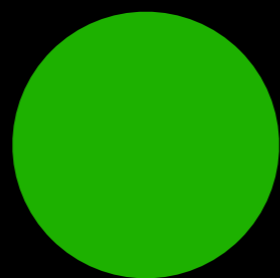
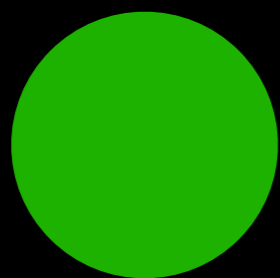


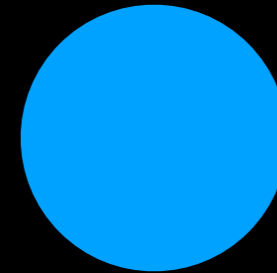
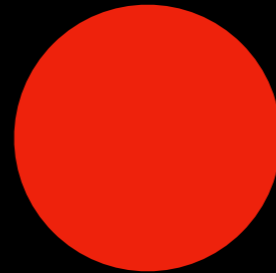
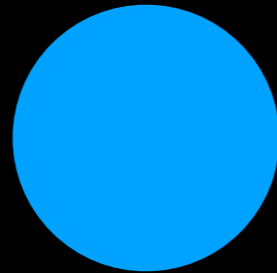
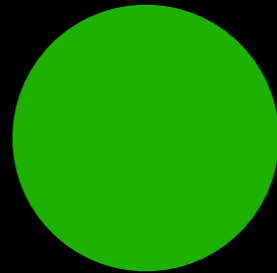
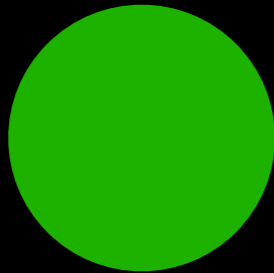




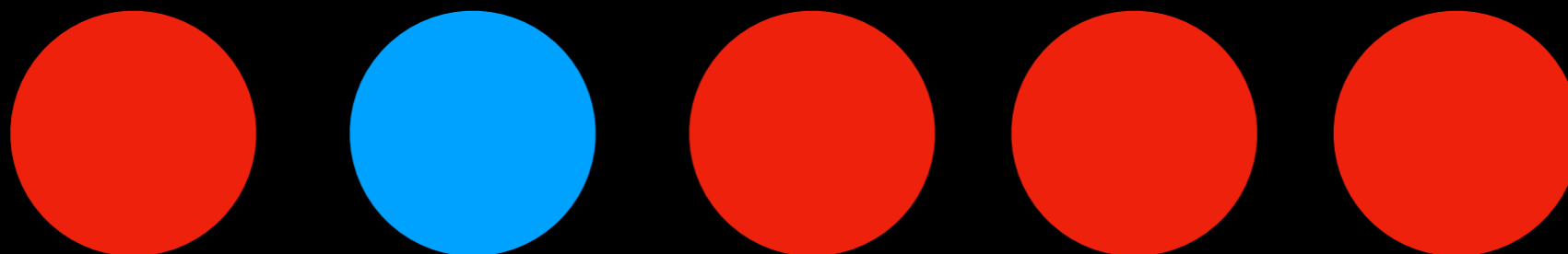


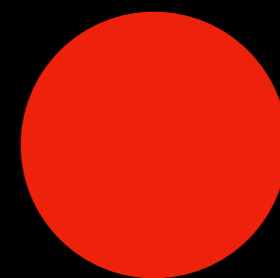
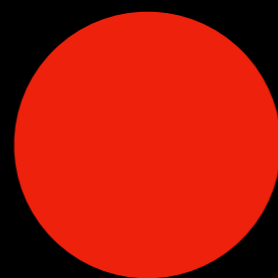
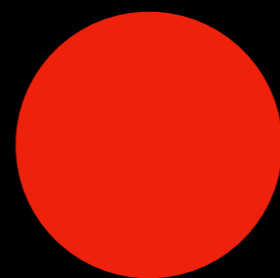
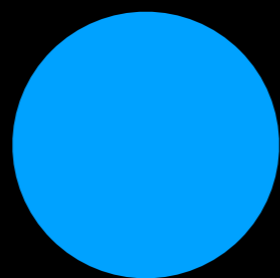
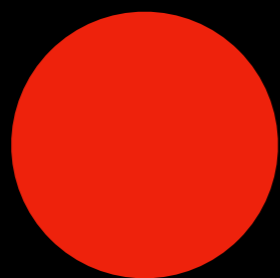




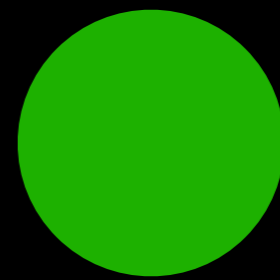
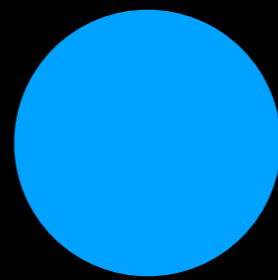
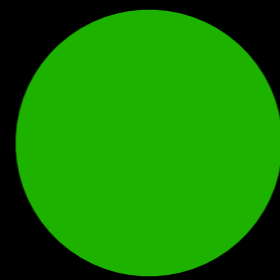
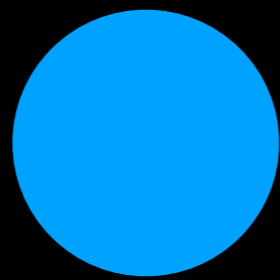
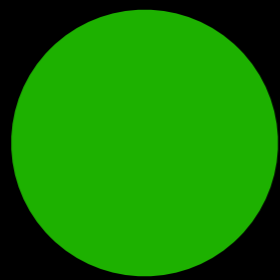


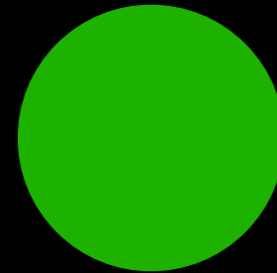
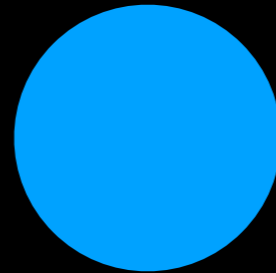
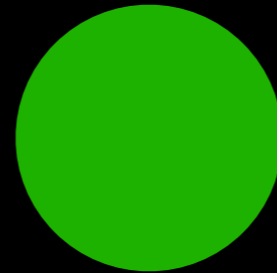
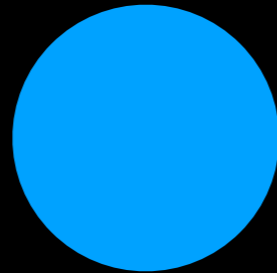
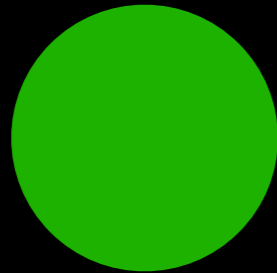










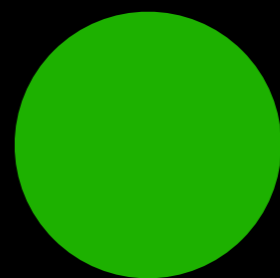
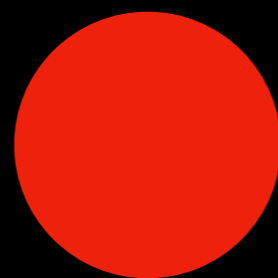
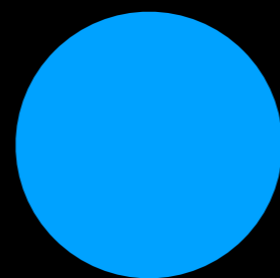
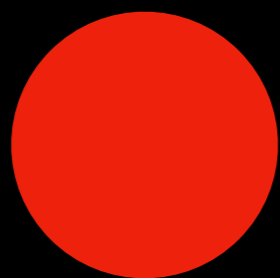


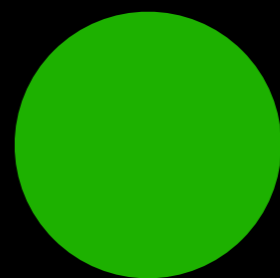
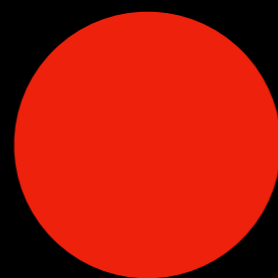
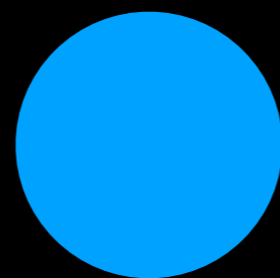
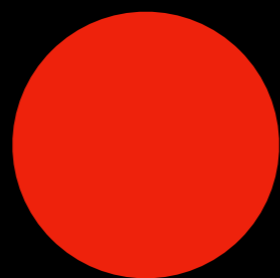
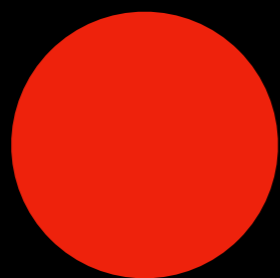




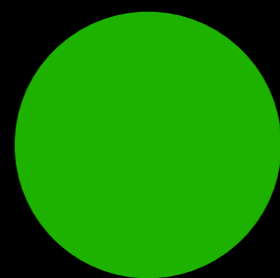
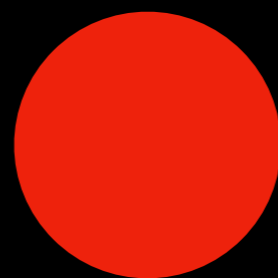
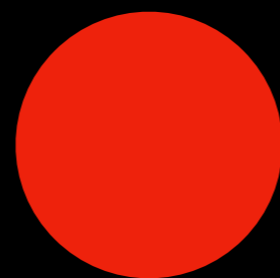
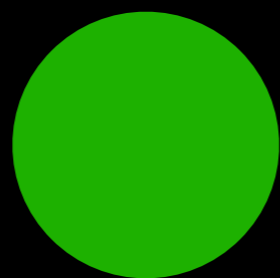
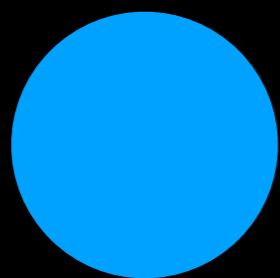
**At most 2 red dots.**

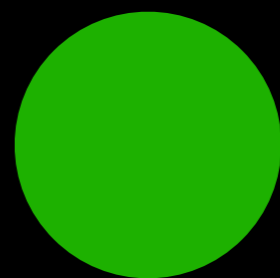
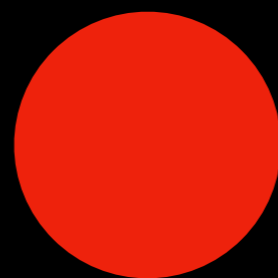
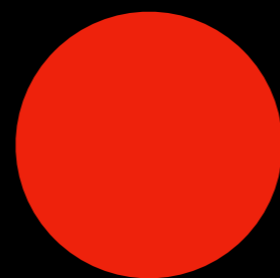
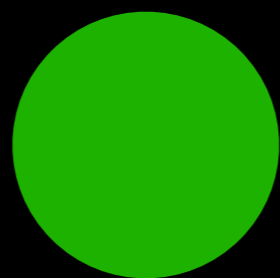
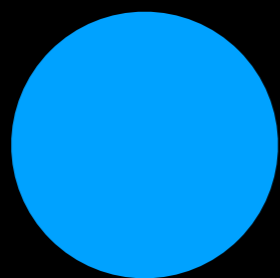






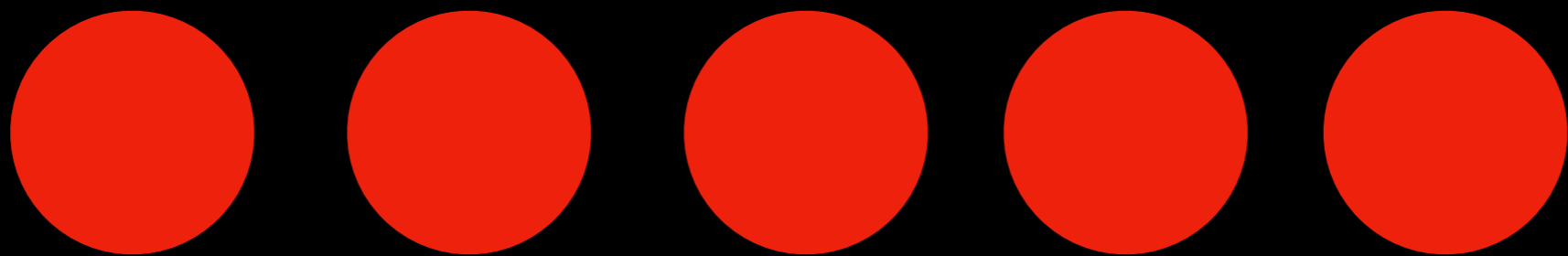


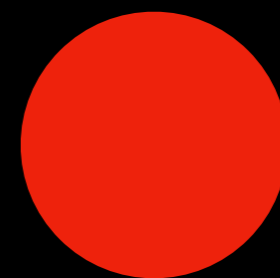
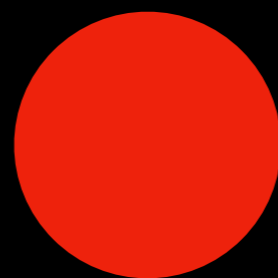
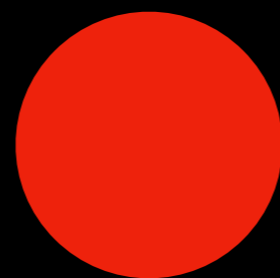
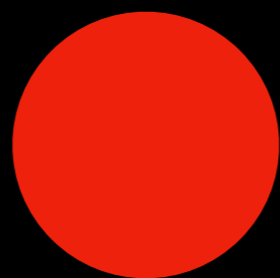
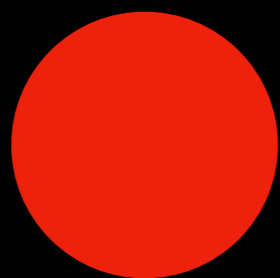




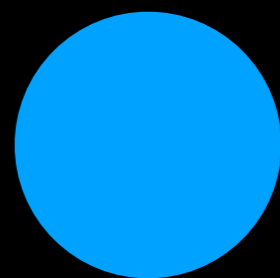
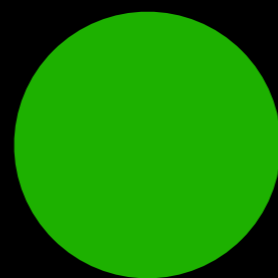
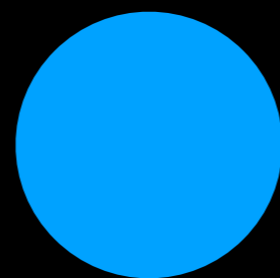
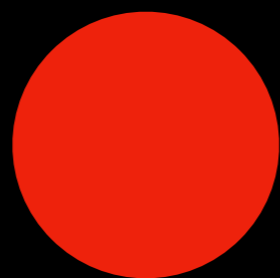
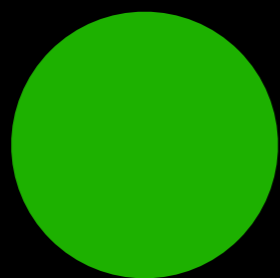


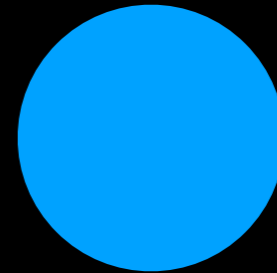
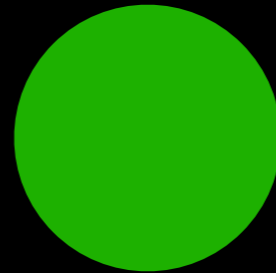
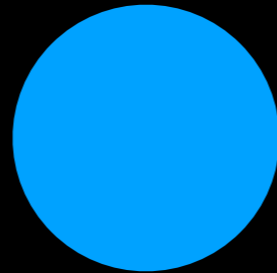
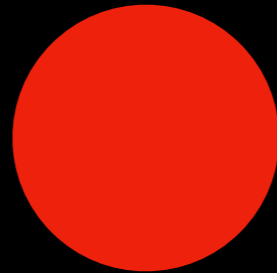
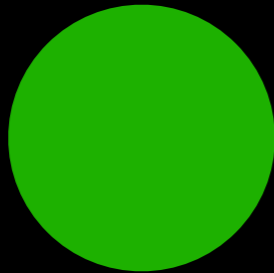




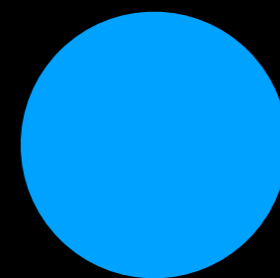
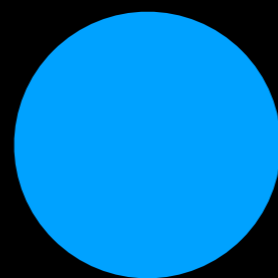
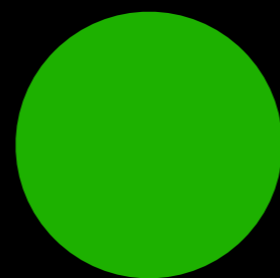
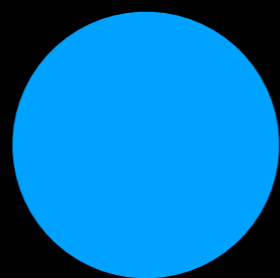
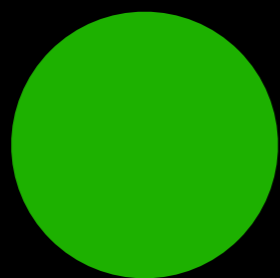




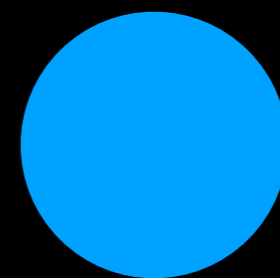
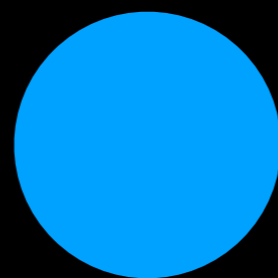
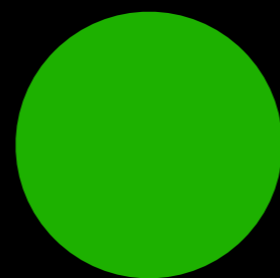
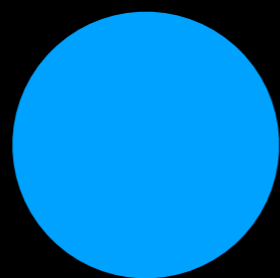
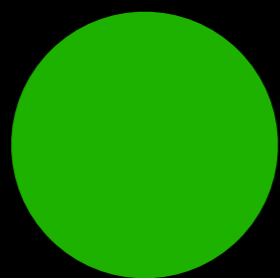




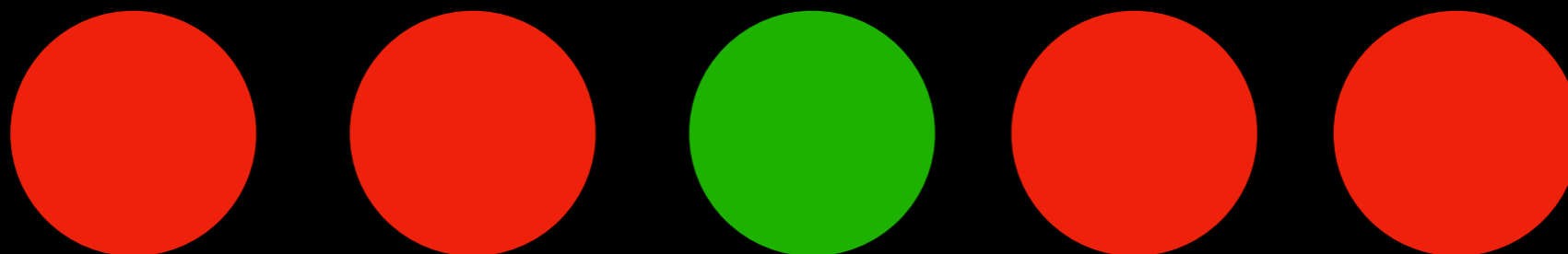


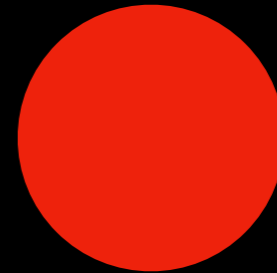
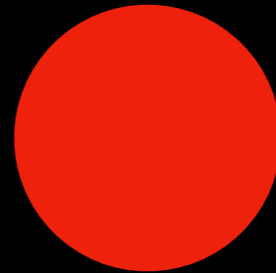
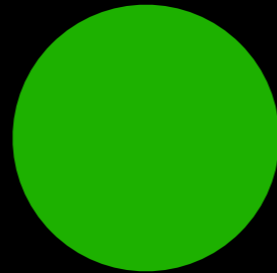
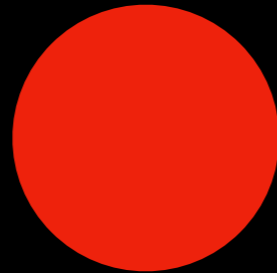
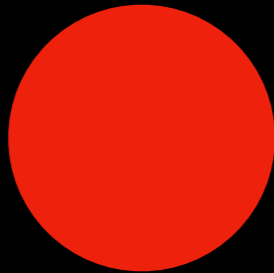














**1, 2, or 4 red dots.**

# Learnability Prediction

**monotone < connected < non-connected**

---

**Condition**

---

**Rules**

Monotone

“There are 0, 1, or 2 red circles.”

“There are 3, 4, or 5 red circles.”

Connected

“There are 1, 2, or 3 red circles.”

“There are 2, 3, or 4 red circles.”

Non-connected

“There are 0, 1, or 5 red circles.”

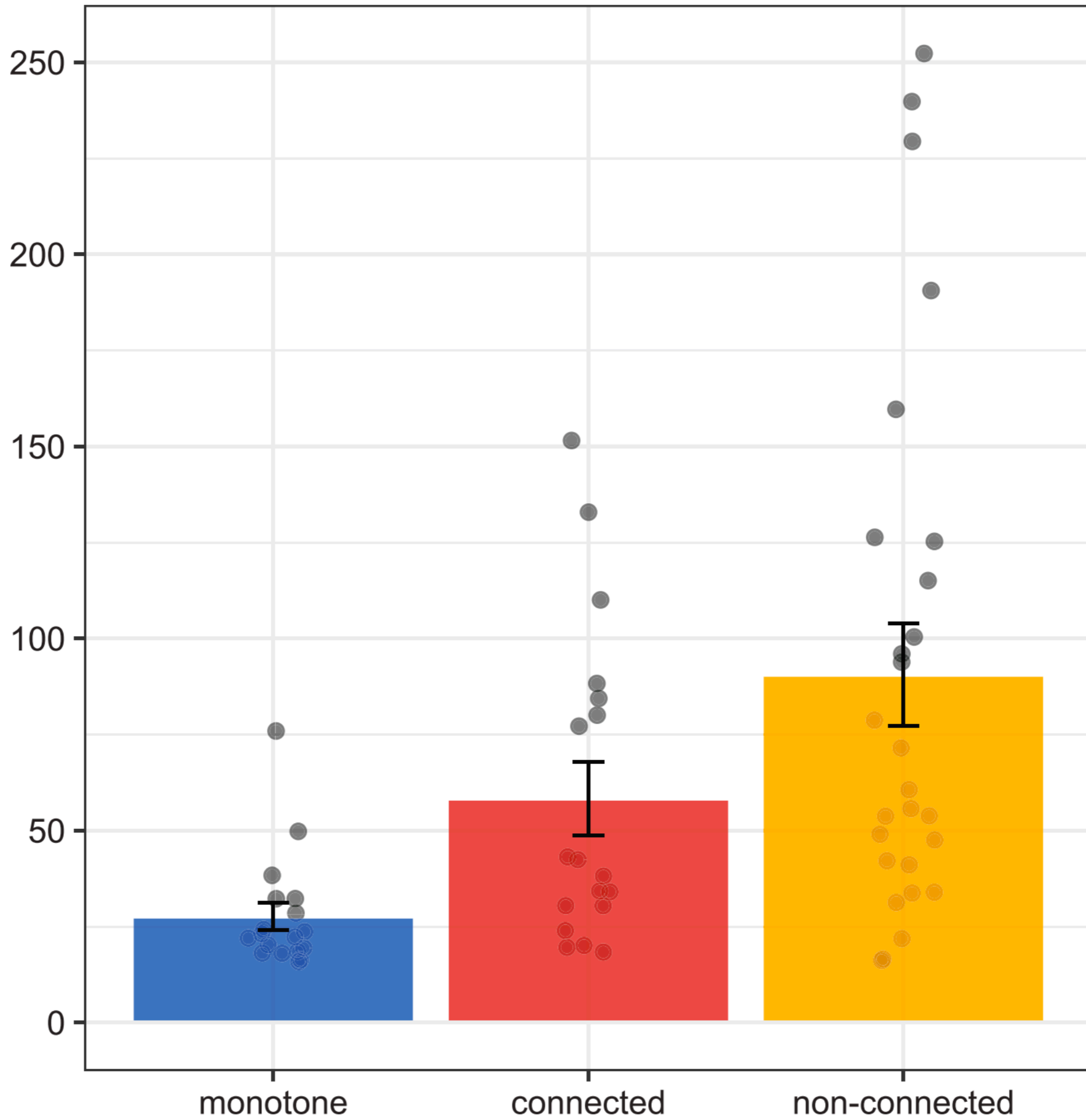
“There are 0, 4, or 5 red circles.”

“There are 1, 2, or 4 red circles.”

---



Nb of Trials needed to reach the learning criterion



# Subtleties

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- Only monotone vs. non-connected was significant

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- “0, 4, or 5” and “0, 1, or 5” much faster than “1, 2, or 4”

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- Only monotone vs. non-connected was significant
- “0, 4, or 5” and “0, 1, or 5” much faster than “1, 2, or 4”
- Connected re-coded as zero, one, or both (rule and negation) *is* then significantly different

# Dynamic Analysis

- Odds of saying “yes” to  $n$  red dots, if already said “yes” to  $n-1$  and  $n+1$  in the same block significantly higher than if for only one or for neither.
- (even controlling for whether the actual rule is connected)

**“The grand goal is to find a list of properties which are, in some sense, double universals: universals across languages, but also across word types....”**

# Constraints on the lexicons of human languages have cognitive roots present in baboons (*Papio papio*)

Emmanuel Chemla<sup>a</sup>, Isabelle Dautriche<sup>b,c,d</sup>, Brian Buccola<sup>a,1,2</sup>, and Joël Fagot<sup>c,d</sup>

<sup>a</sup>Laboratoire de Sciences Cognitives et Psycholinguistique, Département d'Études Cognitives, École Normale Supérieure, PSL University, École des Hautes Études en Sciences Sociales, CNRS, 75005 Paris, France; <sup>b</sup>Centre for Language Evolution, University of Edinburgh, Edinburgh EH8 9YL, United Kingdom; <sup>c</sup>Laboratory of Cognitive Psychology, Aix-Marseille University, CNRS, 13331 Marseille Cedex 3, France; and <sup>d</sup>Institute of Language, Communication, and the Brain, Aix-Marseille University, CNRS, 13604 Aix-en-Provence Cedex 1, France

Edited by Barbara H. Partee, University of Massachusetts Amherst, Amherst, MA, and approved June 14, 2019 (received for review April 24, 2019)

**Using a pattern extraction task, we show that baboons, like humans, have a learning bias that helps them discover connected patterns more easily than disconnected ones—i.e., they favor rules like “contains between 40% and 80% red” over rules like “contains around 30% red or 100% red.” The task was made as similar as possible to a task previously run on humans, which was argued to reveal a bias that is responsible for shaping the lexicons of human languages, both content words (nouns and adjectives) and logical words (quantifiers). The current baboon result thus suggests that the cognitive roots responsible for regularities across the content and logical lexicons of human languages are present in a similar form in other species.**

connectedness | human languages and their lexicons | primate semantics

**H**umans and animals categorize objects in the world into

























show that humans have corresponding learning biases favoring connected quantifiers, as evidenced by performance on rule learning, or pattern extraction, tasks: It is easier to discover connected rules than nonconnected ones, and easier still to discover monotone ones.

A natural hypothesis is that the source of the regularity of the world's lexicons, for both content and logical words, is a learning bias for connectedness. Indeed, Chemla et al. (4) argue that their experimental results with humans support this

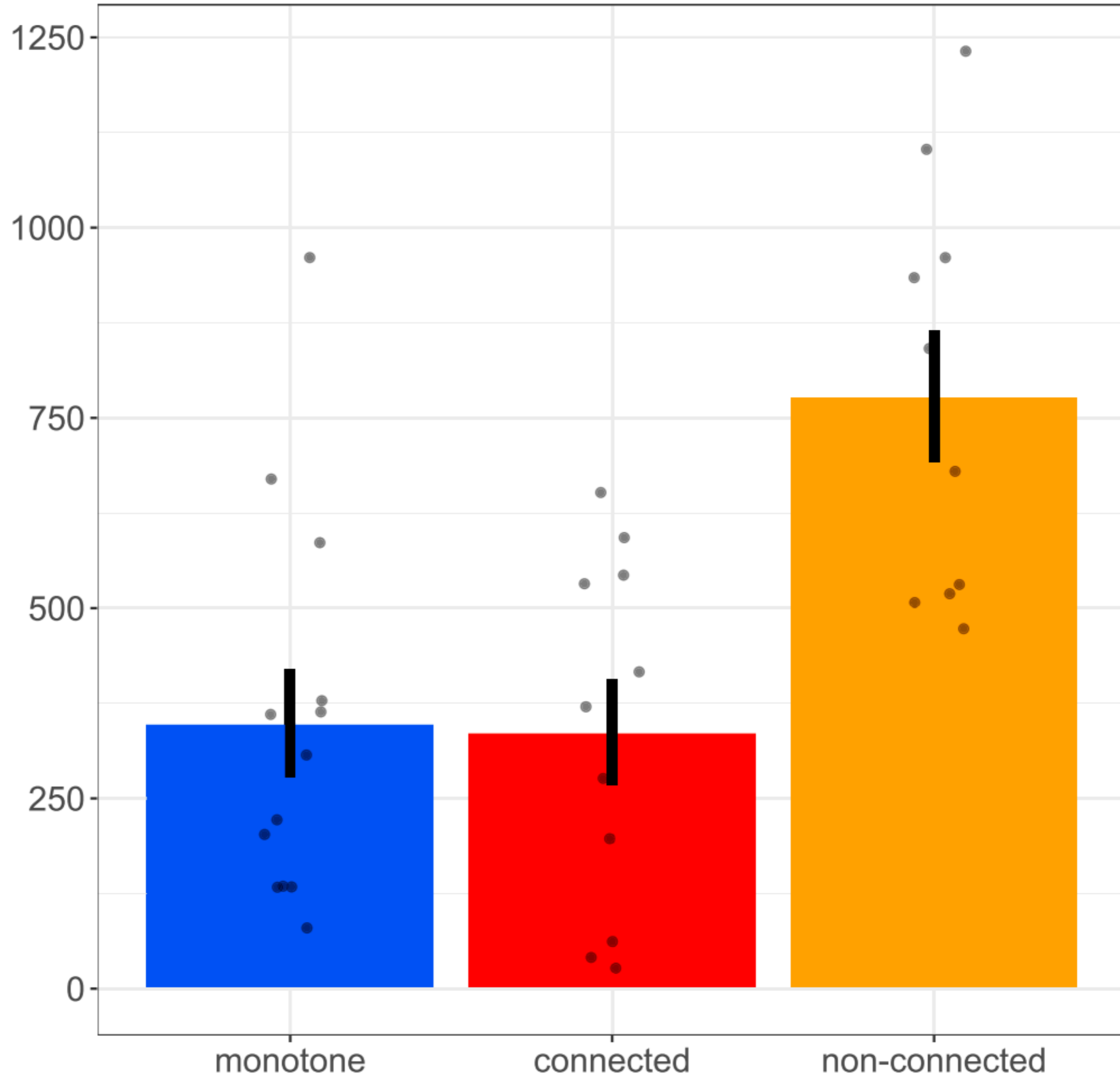
## Significance

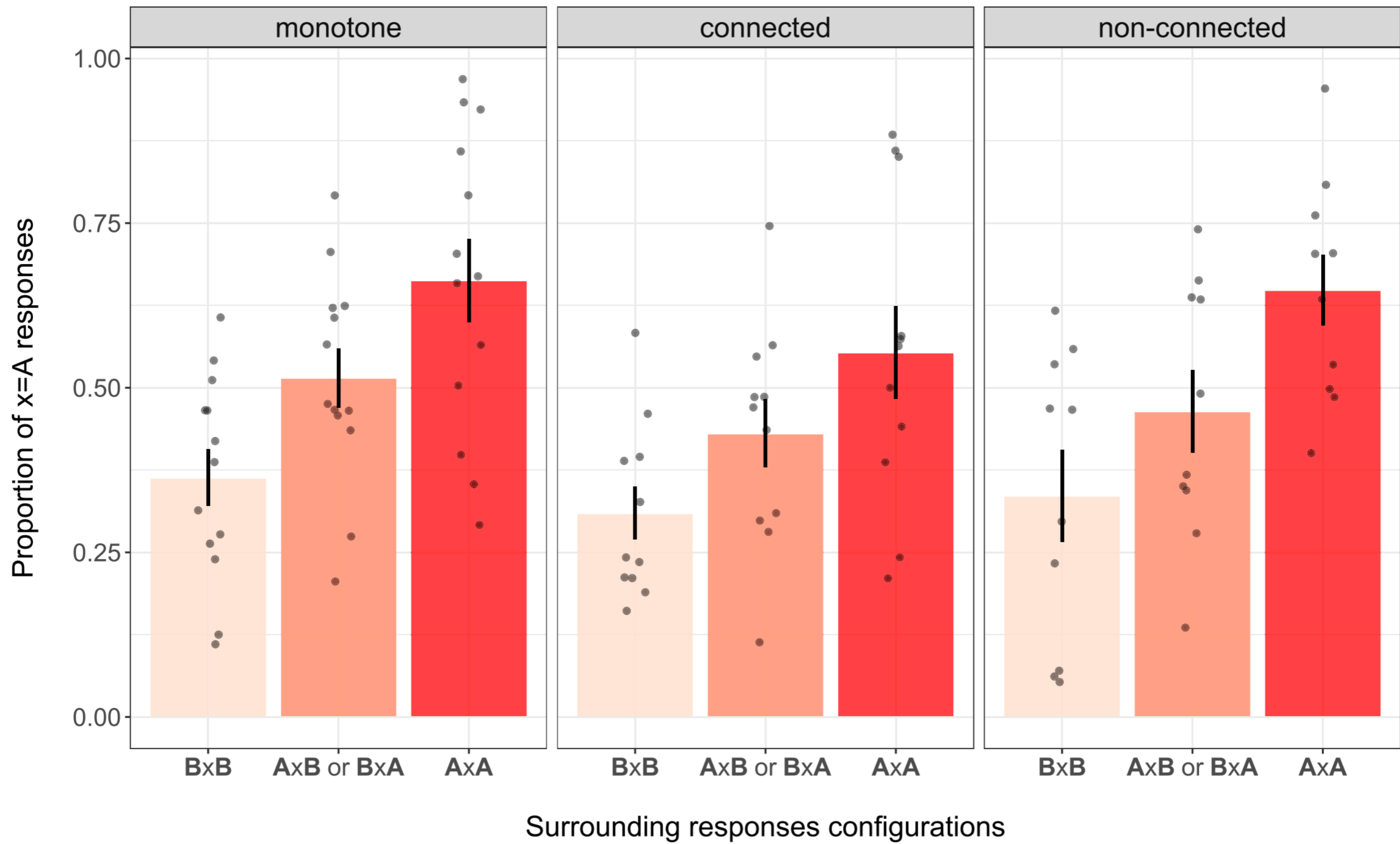
**Universals in language are hard to come by, yet one candidate is that words across the lexicons of the world's languages are, by and large, connected: When a word applies to two**



	<b>Displays</b>						<b>Response buttons</b> (comparison stimuli)	
	0%	20%	40%	60%	80%	100%		
<b>Set 1</b>								
<b>Set 2</b>								
<b>Set 3</b>								

Nb of blocks needed to reach the learning criterion



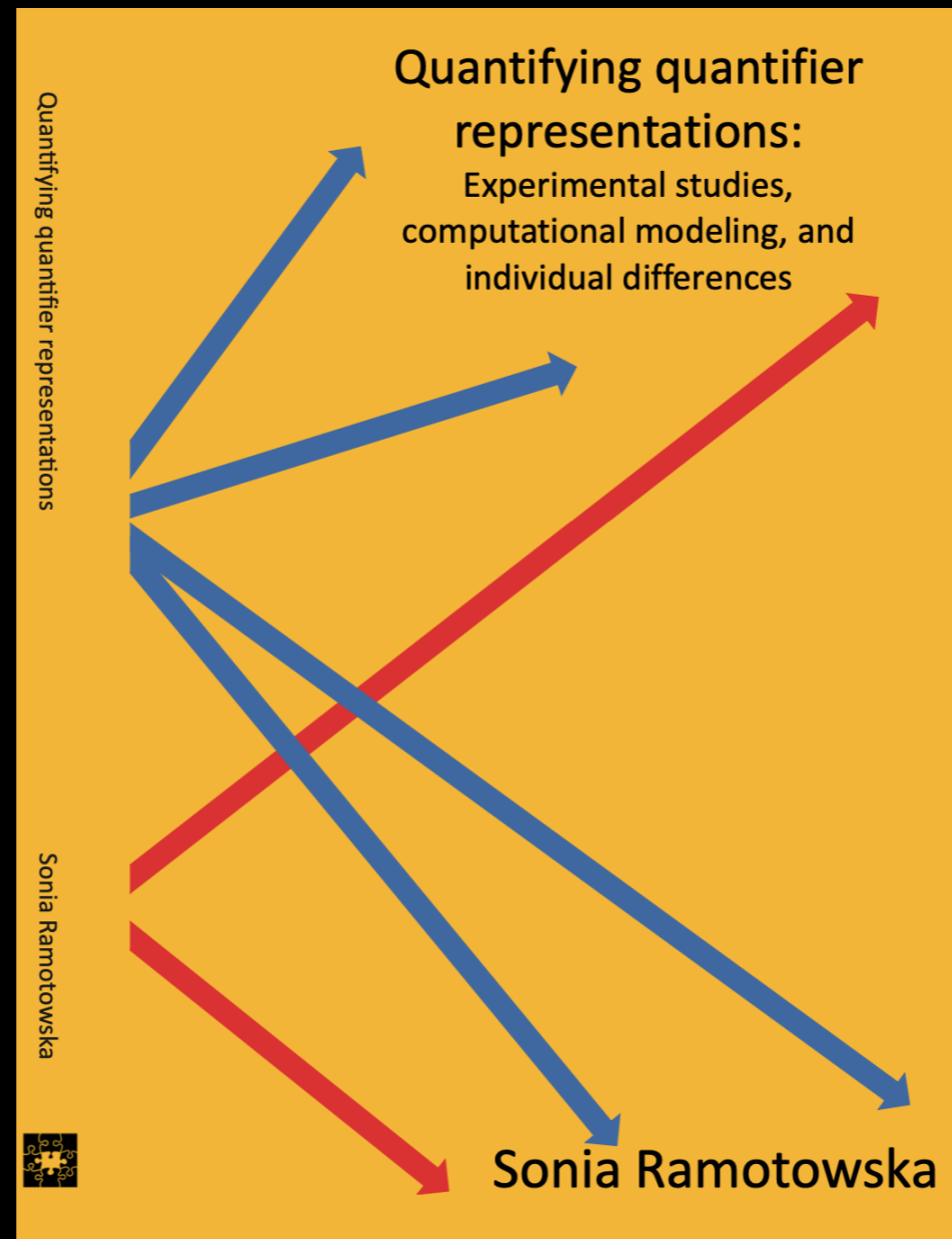


**“The connectedness constraint is thus active in [humans and baboons] in a form that can explain how the referential and functional lexicons of human languages are shaped.”**

# Discussion

- Very small-scale
- No linguistic prompt (e.g. “gleeb of the dots are red”)
- Connectedness vs. monotonicity?

# Large Scale Learnability Experiment

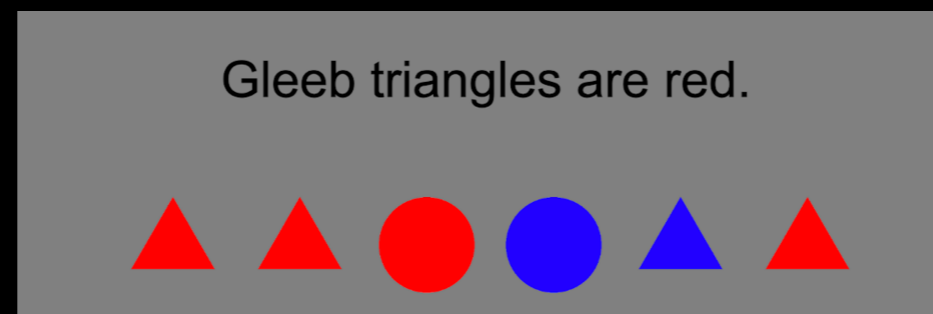
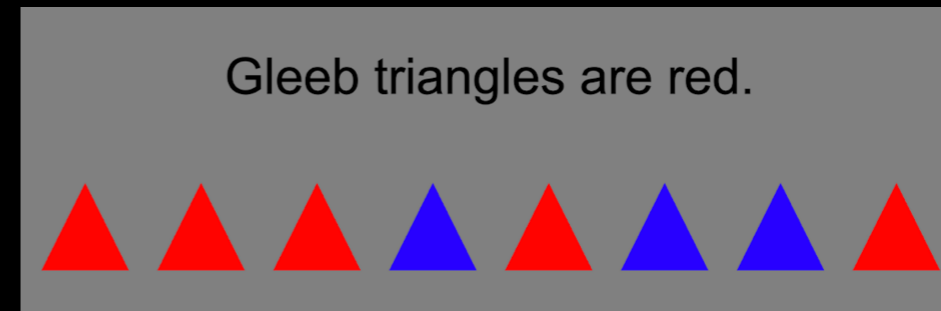
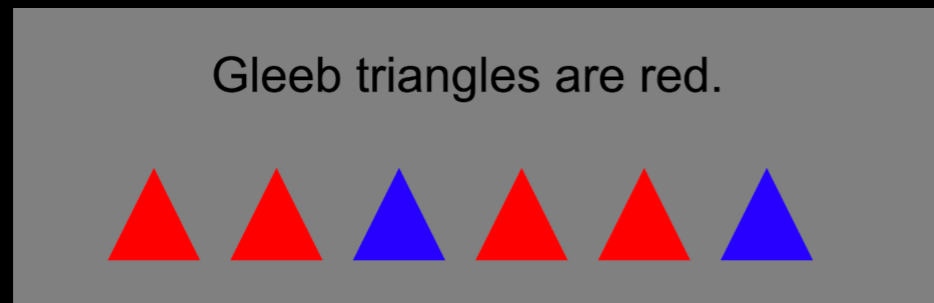


# Tested Quantifiers

- at least 3 & at most 2 vs. between 3 and 6 & at most 2 or at least 7
- between 3 and 6 vs. at most 2 or at least 7
- at least 3 & at most 3 vs. first and the last 3
- not all vs. not only

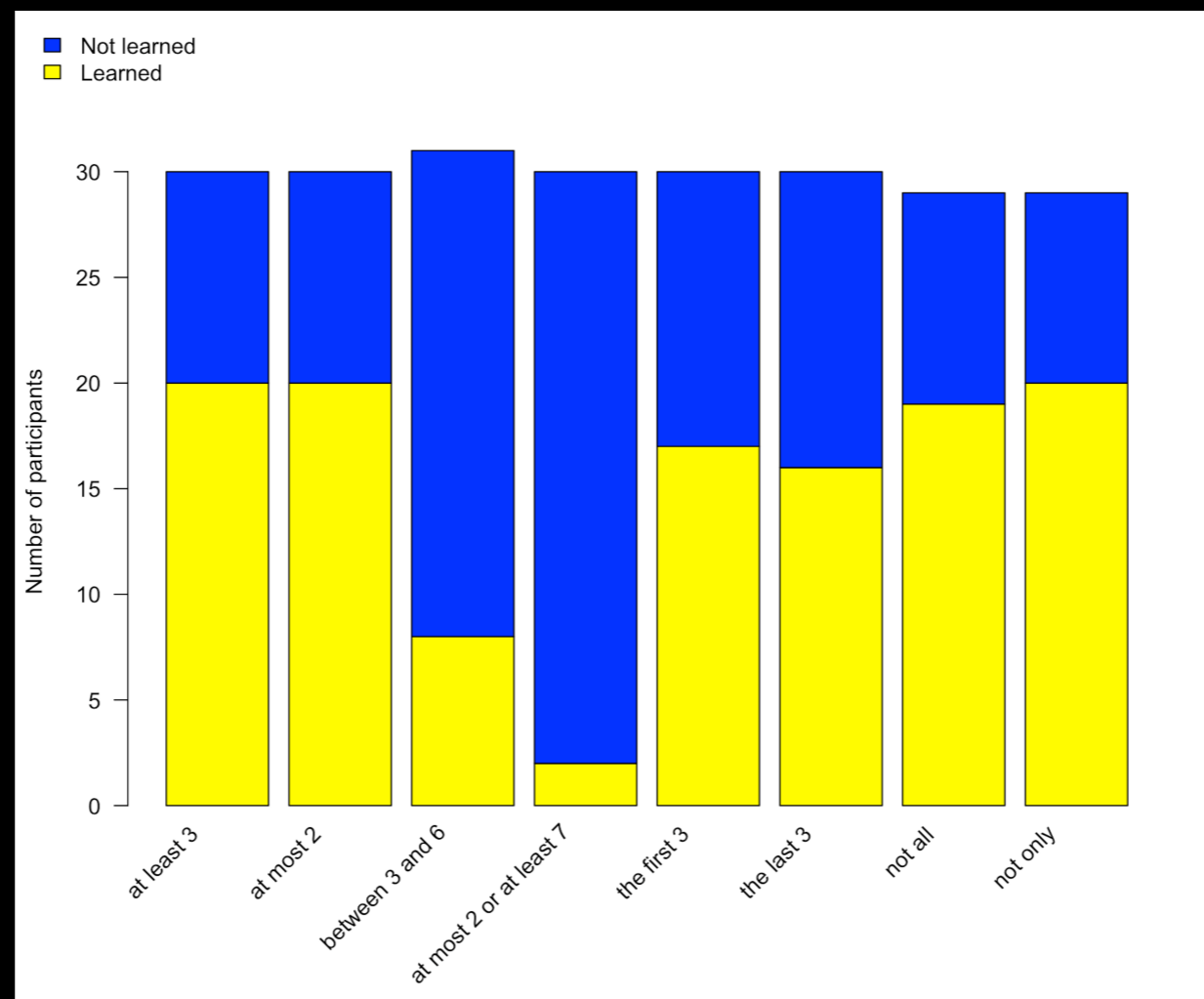
# Design

- ~30 participants for each quantifier (H&L:10, S&dV:9)
- 96 trials, 8 implicit blocks for 12 trials.

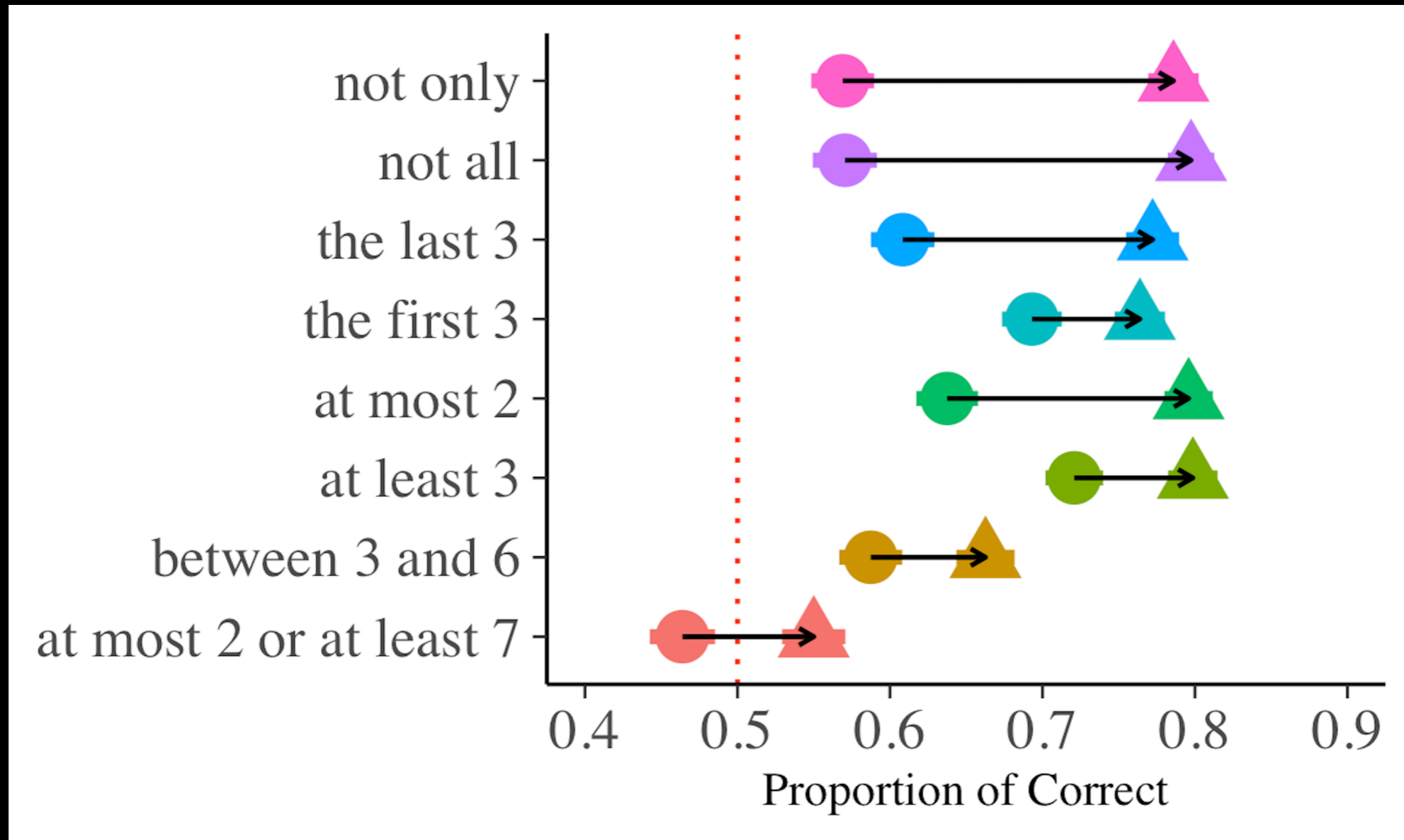




# #participants performing above chance in the last block



# Mean accuracies in the first 25% and the last 25% of the trials for each quantifier



# Mean accuracies in the first 25% and the last 25% of the trials for each universal

