



“Instead of trying to produce a programme to simulate the adult mind, why not rather try to produce one which simulates the child's?”
Alan Turing, 1950.

AI Learning

- Deep learning, reinforcement learning, Supervised learning
- Needs lots of data
- Not much (or right) generalization
- Computationally tractable

2-year-olds' learning

- Very little data
- Excellent generalizations
- Search and sampling
- Computationally intractable

DARPA Machine Common Sense: MESS Model-building, exploratory, social learning systems

- Abstract causal models from statistical evidence
- Active learning through exploratory play
- Social learning through imitation and testimony

Probabilistic Causal Models in Children (Pearl, 2000, Spirtes et al. 2001, Gopnik & Wellman, 2012, Gopnik 2020)

Four-year-olds (and younger) can rationally

- Infer complex causal structure (chains versus common effects vs common causes) from conditional probabilities (Schulz et al. 2007)
- Integrate and override prior causal knowledge in the face of new evidence (Kushnir & Gopnik, 2007, Griffiths et al. 2011)
- Infer unobserved causal structure (Gopnik et al. 2004)
- Infer causal theories of the physical, biological, psychological and social domains (Schulz & Gopnik, 2004, Kushnir et al. 2010, Seiver et al. 2013, Vasilyeva et al. 2018)
- Infer and use counterfactuals (Buchsbaum et al. 2012)
- Infer abstract over-hypotheses (Lucas et al. 2014, Gopnik et al. 2017)

Variable Selection and Analogical Reasoning

- M. Goddu, & A. Gopnik. (2020) Learning what to change: Young children use 'difference-making' to identify causally relevant variables. *Developmental Psychology*, 56, 2, 275
DOI:10.1037/dev0000872
- M. Goddu, T. Lombrozo, & A. Gopnik. (2020). Transformations and transfer: Preschool children understand abstract relations and reason analogically in a causal task. *Child Development*. 91, 6, 1898-1915, *DOI: 10.1111/cdev.13412*



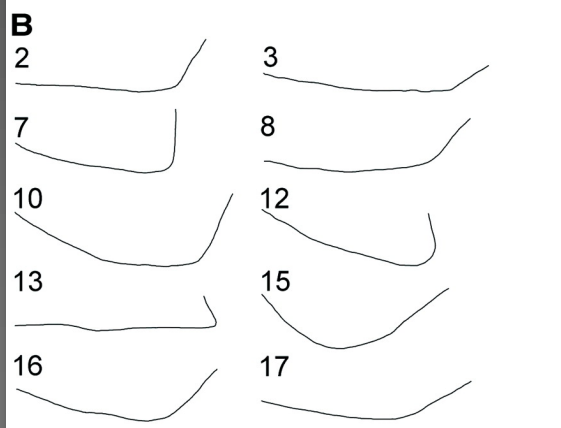
Intervention, Exploration and Active Learning

- Schulz et al. 2007
- Schulz & Bonawitz, 2007
- Bonawitz et al. 2012
- Ruggieri et al. 2015, 2019



The Explore-Exploit Problem

Longer Childhood, Larger Brain, Smarter Animal



Hypothesis

Childhood is evolution's way of resolving explore/exploit trade-offs and performing simulated annealing.

Gopnik et al. 2017, PNAS, Gopnik,
Philosophical Transactions of the Royal
Society B, 2020

Explore Features, Exploit Bugs

- Noisiness, variability, randomness
- Risk-taking
- Impulsivity
- Play
- Curiosity



The Bickert Detector

Liquin & Gopnik: Children are more exploratory and learn more than adults in an approach-avoid task. *Cognition*, Volume 218, January 2022, 104940 *Cognition* 2021



APPROACH
or
AVOID

APPROACH
or
AVOID

APPROACH
or
AVOID

APPROACH
or
AVOID



Actual Outcome:

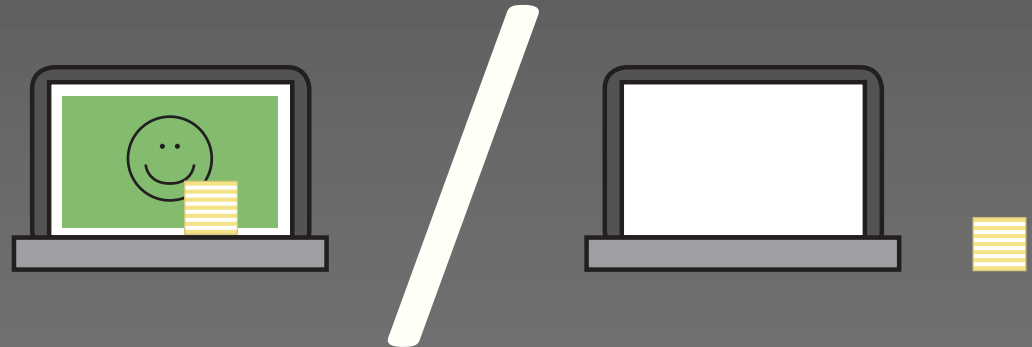


Stickers:





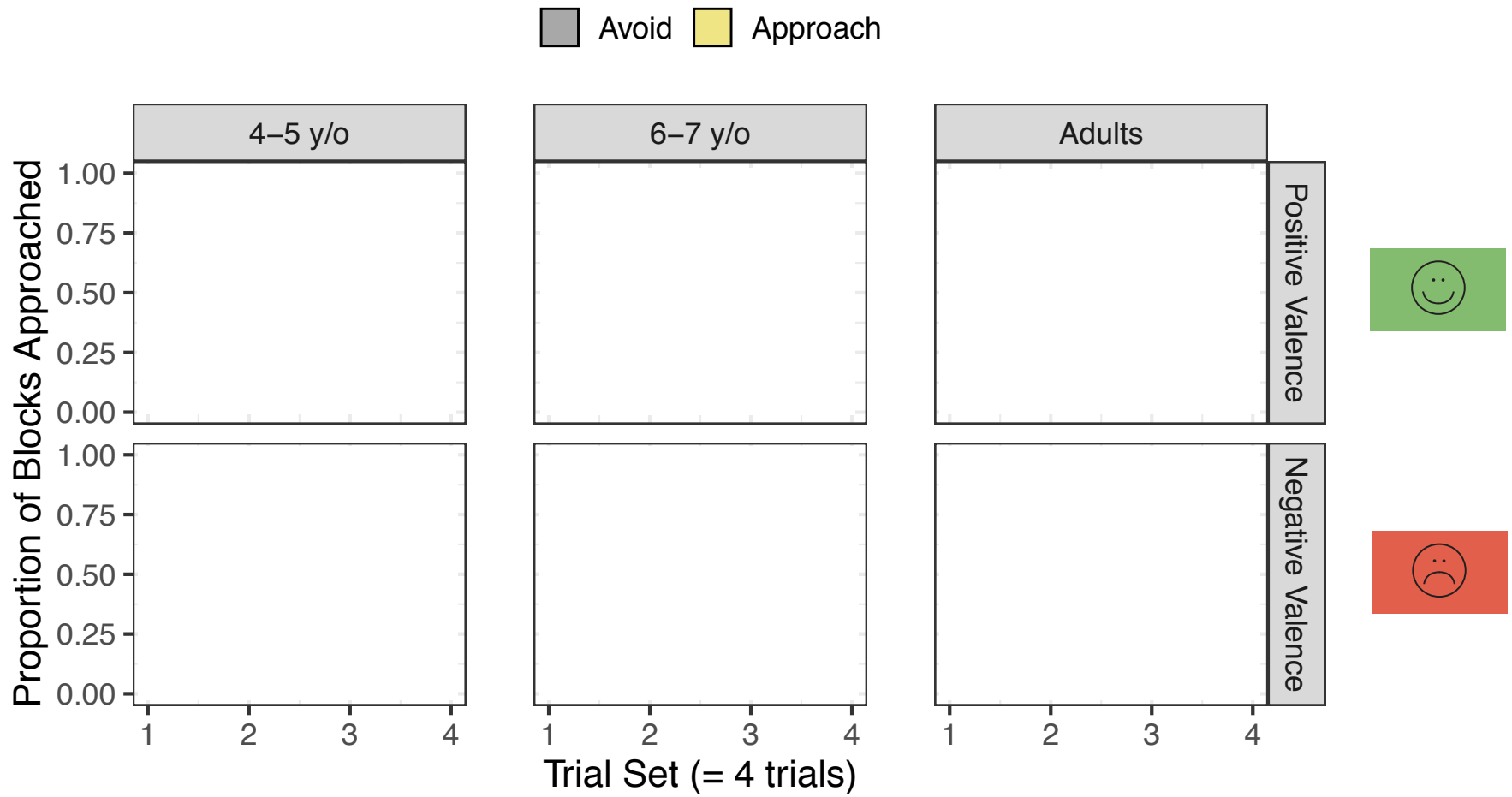
x 4 sets



“Approach”

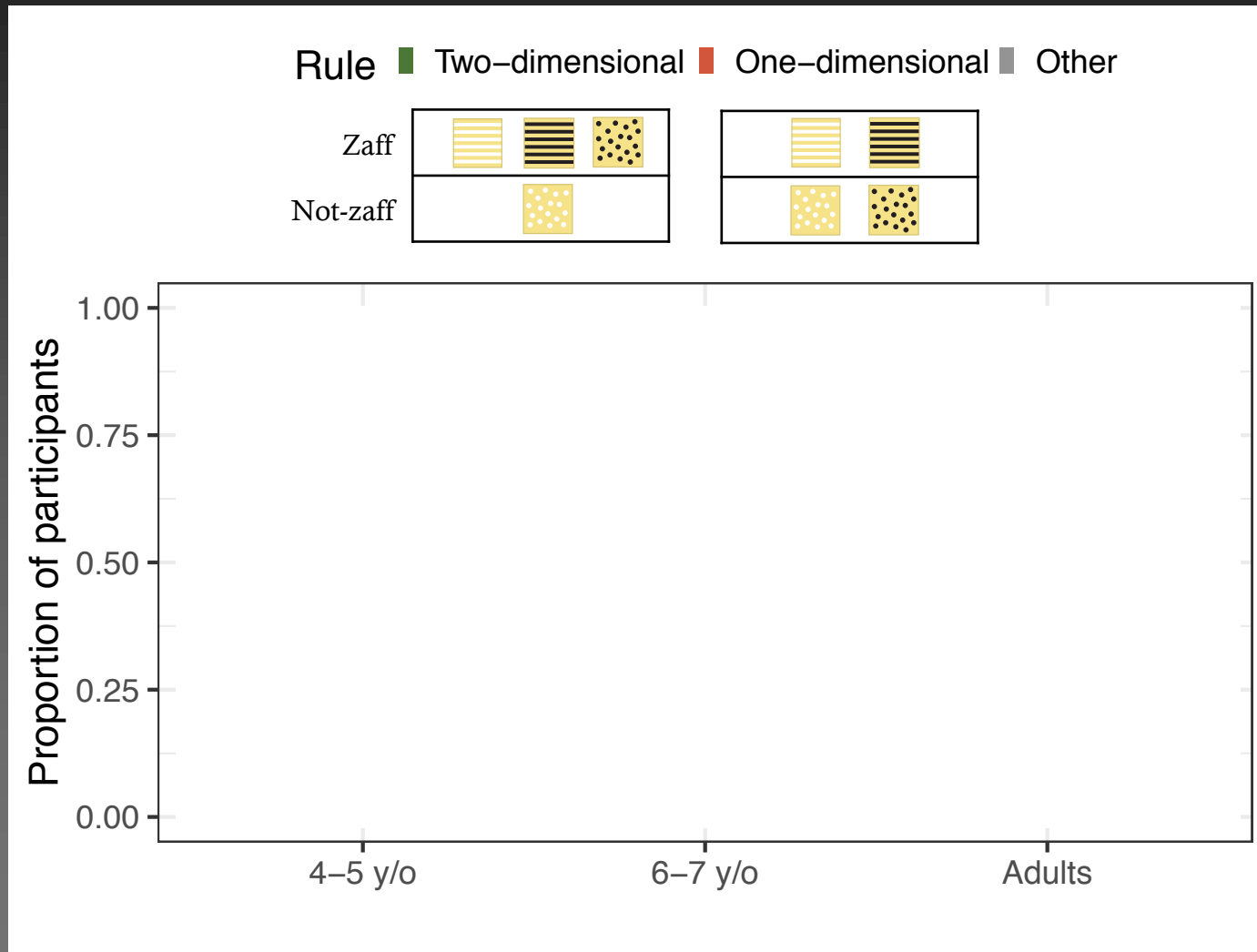
“Avoid”

Approach/Avoid Decisions



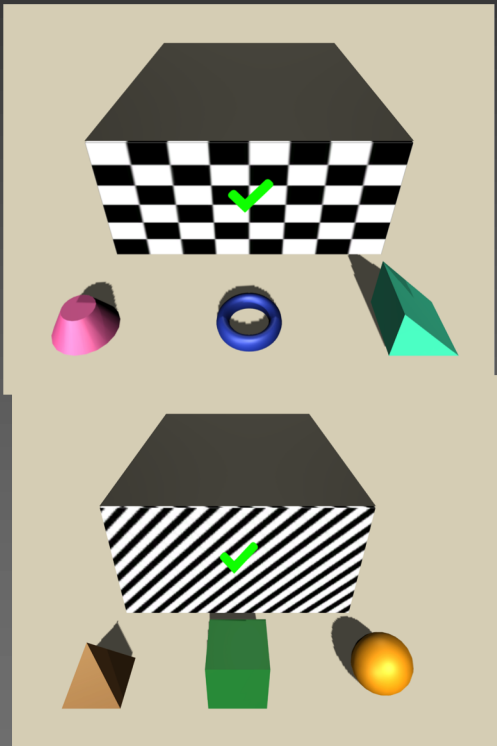
Positive Blocks: Age Group $F(2,106) = 28.41, p < .001^{***}$
Negative Blocks: Age Group $F(2,105) = 14.05, p < .001^{***}$

Is this reflected in learning?



$\chi^2(2) = 17.33, p < .001^{***}$

EXP outline:



GIVEN HYPOTHESIS SPACE

NOT GIVEN HYPOTHESIS SPACE

CONJUNCTIVE

DISJUNCTIVE

A3: Blocks DEF
Blickets are D&E

N: 22

Turns on: DE, DEF

Does not turn on: D,E,F,DF, EF

Predictions: kids will try objects 1 by 1 until they see that non make it go, then do they switch to try combinations of blocks, might do it here since we gave them evidence

A4: Blocks DEF
Blickets are D'or'E

N: 23

Turns on:
D,DE,DEF,E,EF,DF

Does not turn on:F

Predictions: they try objects 1 by 1, see it works, stop there

B3: Blocks DEF
Blickets are D&E

N:20

Turns on: DE,DEF

Does not turn on: D,E,F,DF,EF

Predictions: kids will try objects 1 by 1 until they see that non make it go, then do they switch to try combinations of blocks, might do it here since we gave them evidence

B4: Blocks DEF
Blickets are D 'or' E

N:20

Turns on:D,DE,DEF,E,EF,DF

Does not turn on:F

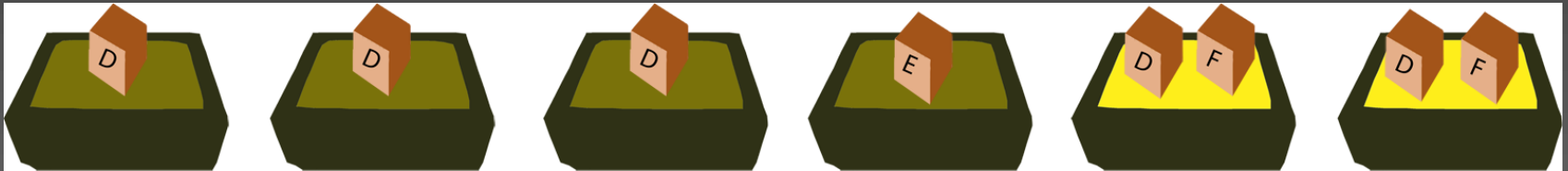
Predictions: they try objects 1 by 1, see it works, stop there

When Younger Learners are More Exploratory

- A. Gopnik, T. Griffiths, & C. Lucas (2015). *Current Directions in Psychological Science*, 24 (2), 87-92
- C. Lucas, S. Bridgers, T. Griffiths, & A. Gopnik (2014). *Cognition*. 131, 2, 284–299.
- A. Gopnik, S. O’Grady, C. Lucas, T. Griffiths A. Wente, S. Bridgers, R. Aboody, H. Fung, R. E. Dahl, (2017). *PNAS*.

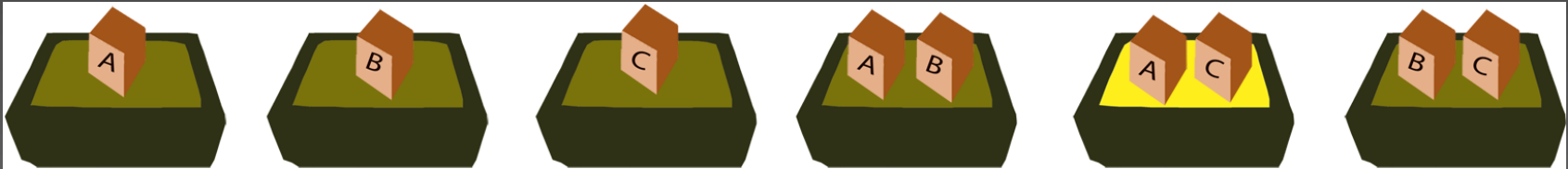


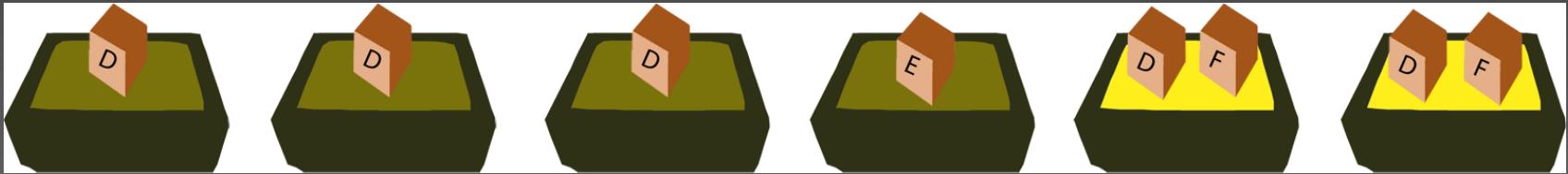
Which objects are blickets?



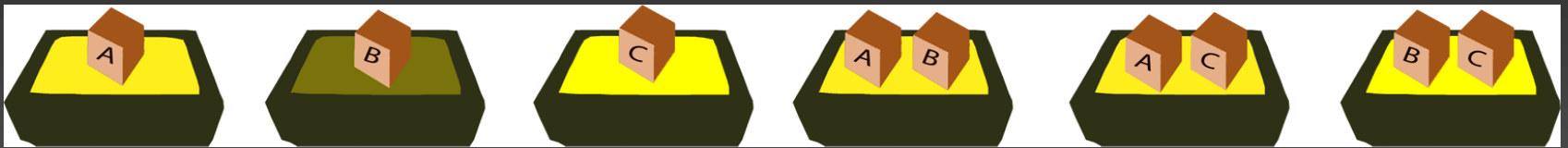
Is D a blicket? Is E a blicket? Is F a blicket?

What if you also saw these events?

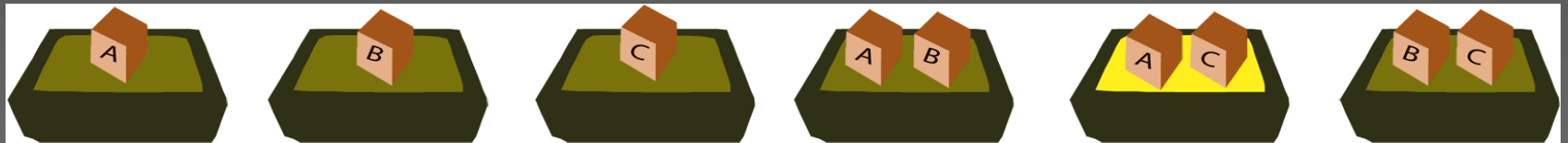




Disjunctive Training



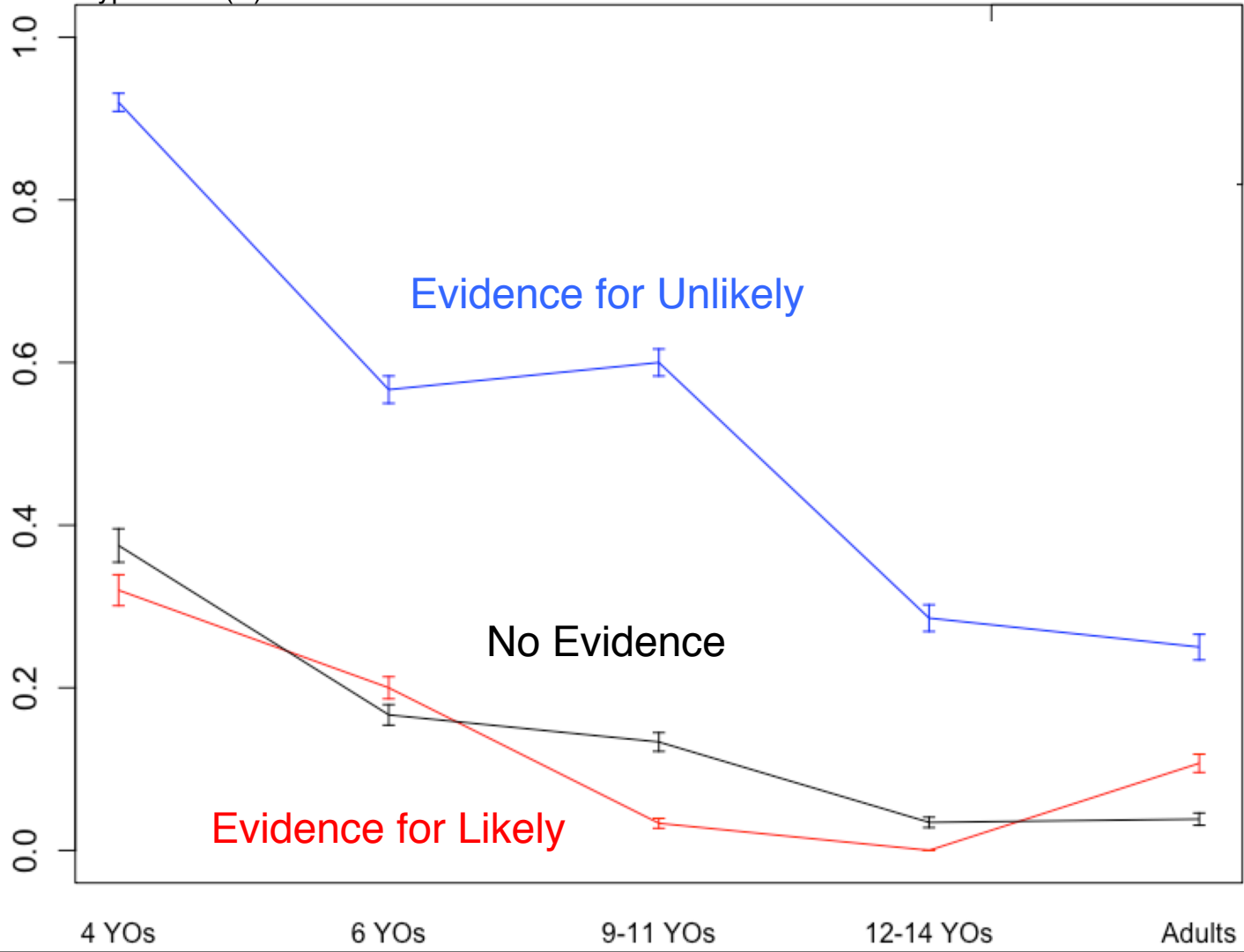
Conjunctive Training



Test

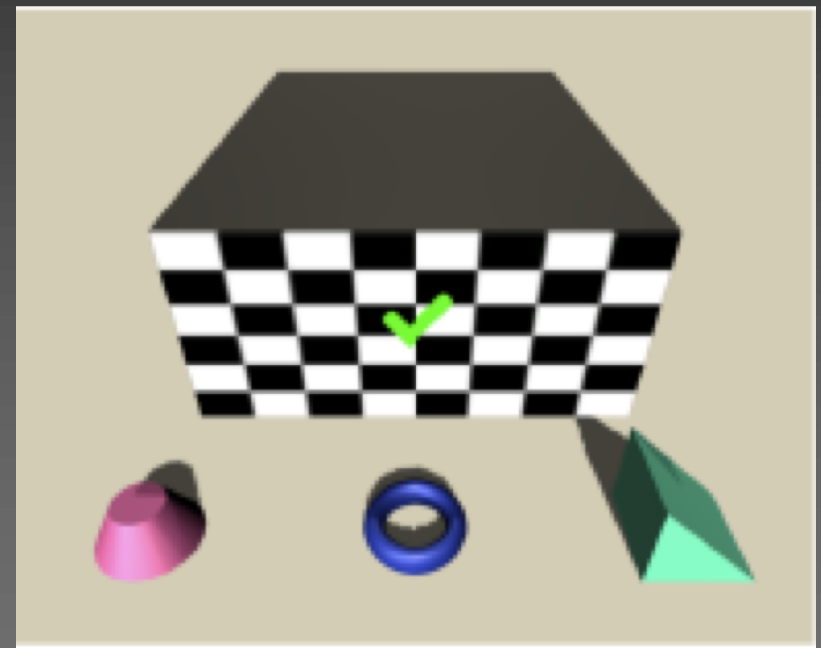


Gopnik et al. PNAS, 2017 Proportion of Participants Choosing the Unlikely Physical Hypothesis (D)



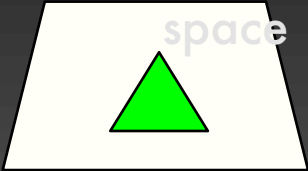
Exploration of Causal Structure

- Learning Casual Overhypotheses through Exploration in Children and Computational Models ,
- Rosemary Ke, Eliza Kosoy, Jessica Hamrick, Jasmine Collins, David Chan, Sandy Huang, Adrian Lu

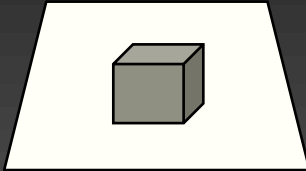


Condition: Not given hypothesis

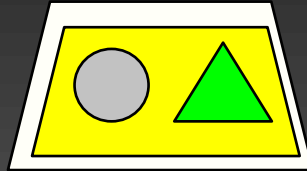
space



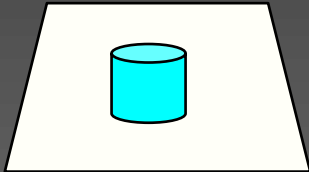
Doesn't turn on



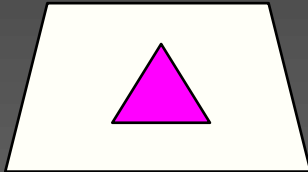
Doesn't turn on



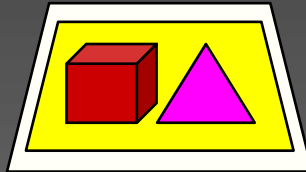
Turns on



Doesn't turn on



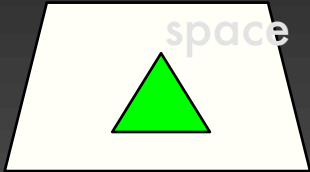
Doesn't turn on



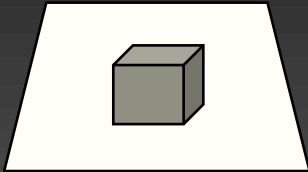
Turns on

Visualizations for paper:

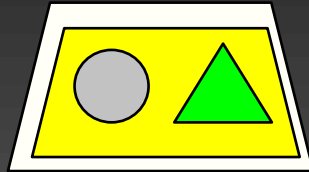
Condition: Given hypothesis



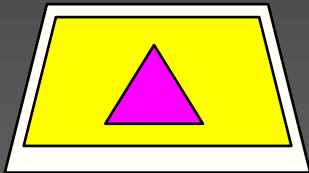
Doesn't turn on



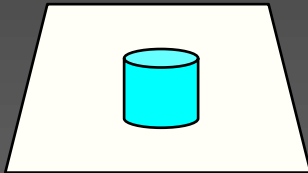
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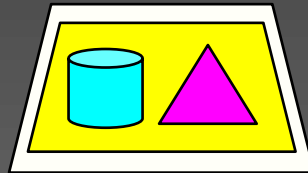
Turns on



Turns on



Doesn't turn on



Turns on

Visualizations for paper:

Collaborators and Support

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NSF

- The McDonnell Foundation Causal Learning Collaborative
- The Bezos Foundation
- DARPA MCS