

DeepMind

Tell me why!

**Explanations support learning relational
and causal structure**

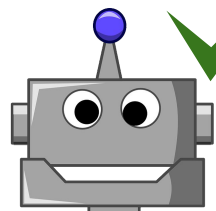
Andrew Lampinen

Nick Roy, Ishita Dasgupta, Stephanie Chan, Allison Tam, Chen
Yan, Jay McClelland, Adam Santoro, Neil Rabinowitz, Jane Wang,
Felix Hill

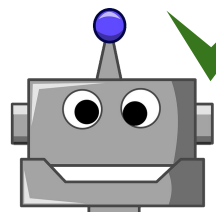


25/2/2022

RL: promises and challenges



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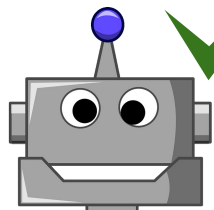
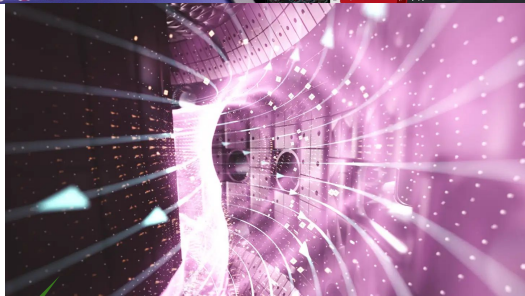


Reward is enough **for intelligence?**

David Silver , Satinder Singh, Doina Precup, Richard S. Sutton

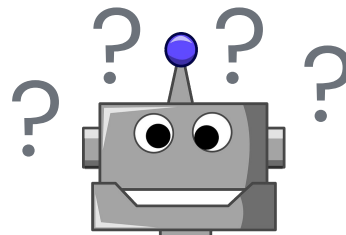
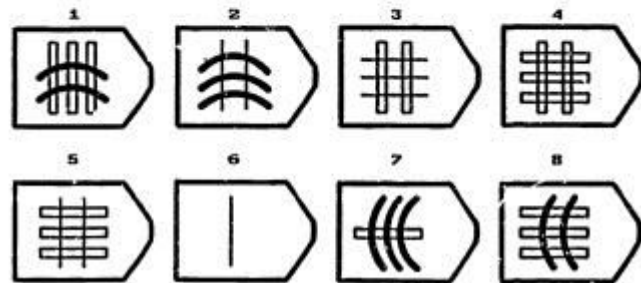
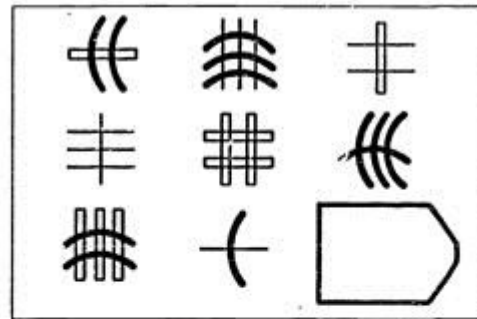


RL: promises and challenges



Reward is enough **for intelligence?**

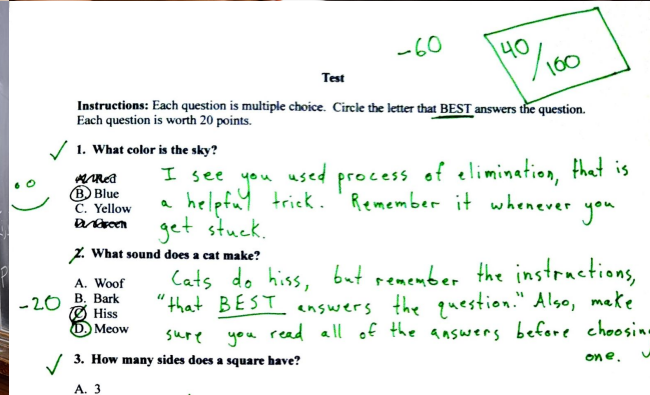
David Silver , Satinder Singh, Doina Precup, Richard S. Sutton



What is different about human learning?



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Explanation along with reward!

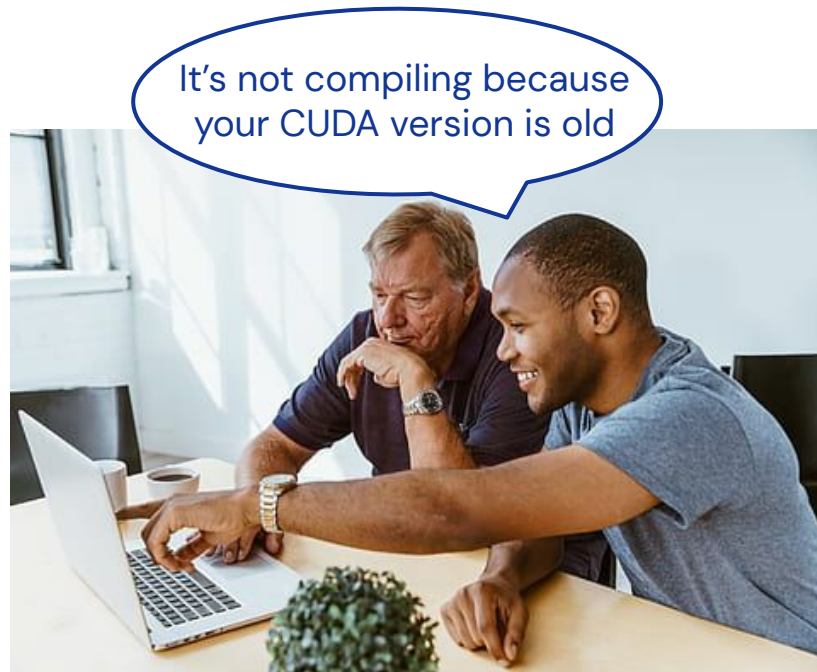


What is an explanation?

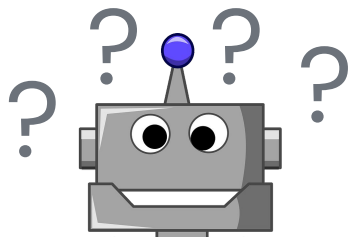
Explanations are intended to *communicate* the links between:

- Concrete situation
- Abstract principles which are:
 - Generalizable
 - Causal

Explanations help us to learn and generalize abstract tasks!



Could agents learn and generalize better if trained with explanations?



Odd-one-out task: abstraction and relational reasoning



(lots of prior work in cog/neuro, e.g. Stephens & Navarro, 2008; Crutch et al., 2009)



Odd-one-out task: abstraction and relational reasoning

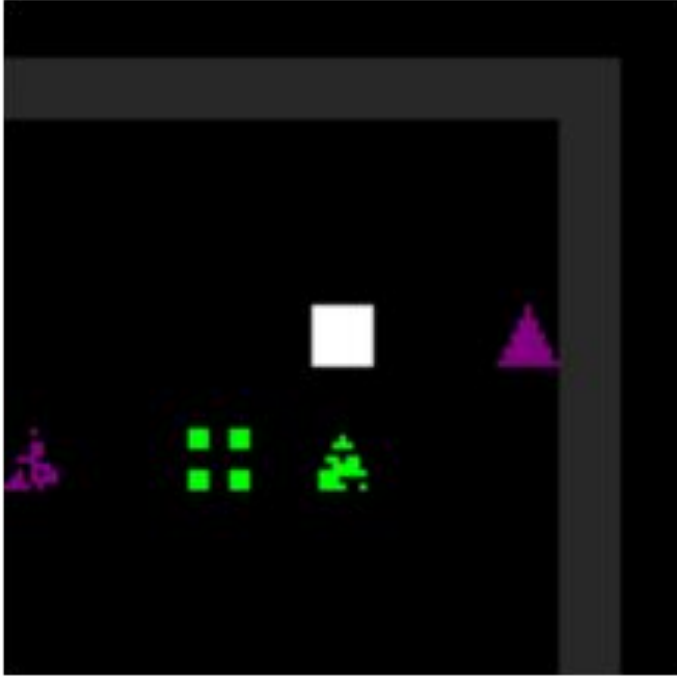


Proper subsets don't reveal the answer!

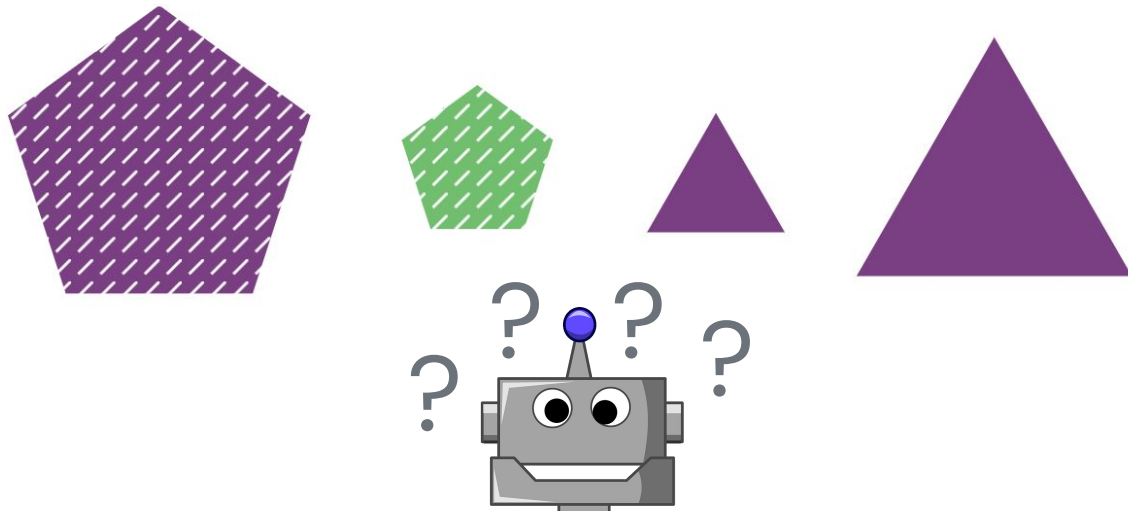
A challenging credit assignment problem from reward alone.



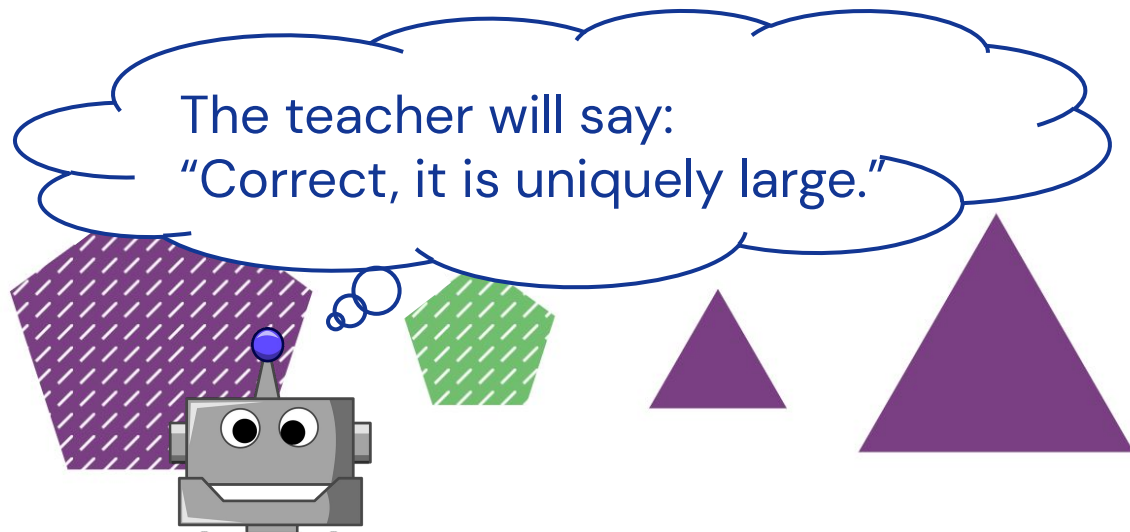
Instantiated in 2D and 3D environments



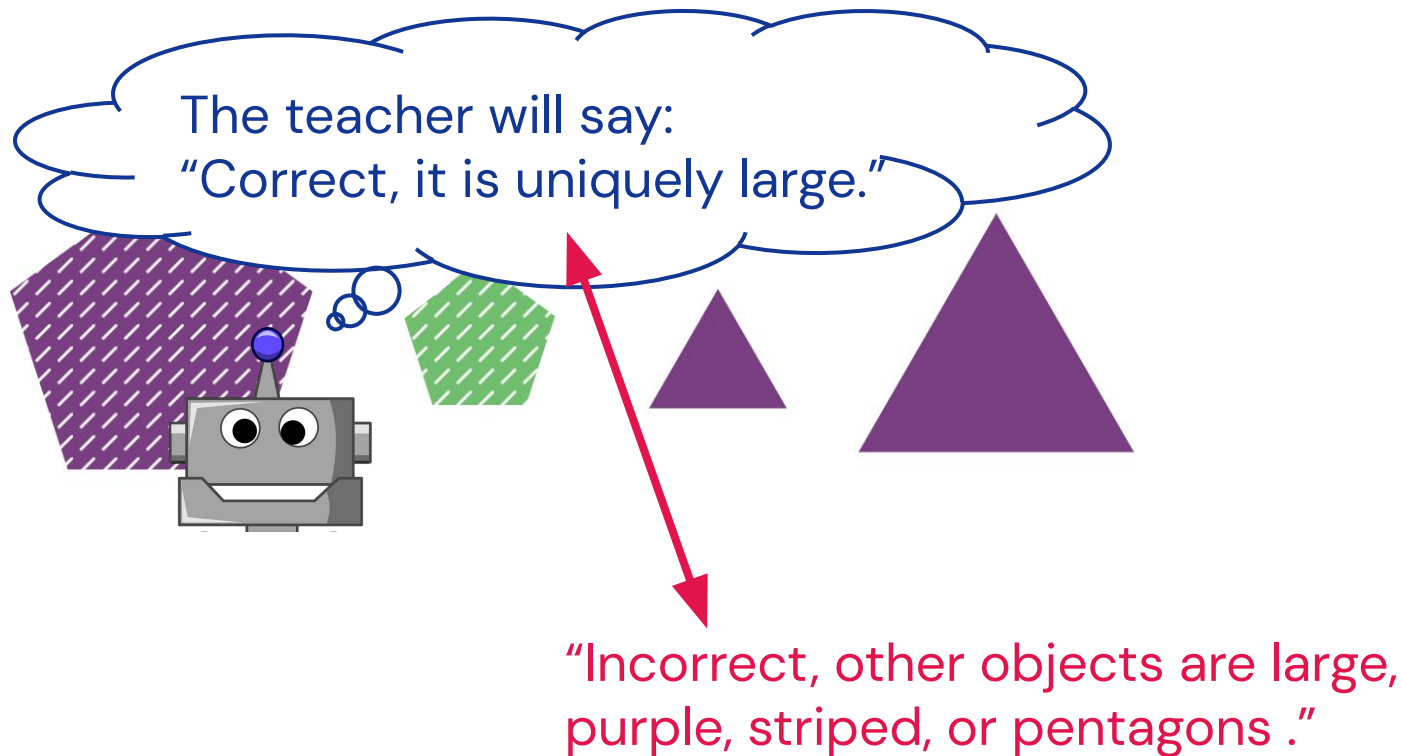
Agents struggle to learn odd-one-out tasks from rewards alone:
abstractions and relations are hard to infer!



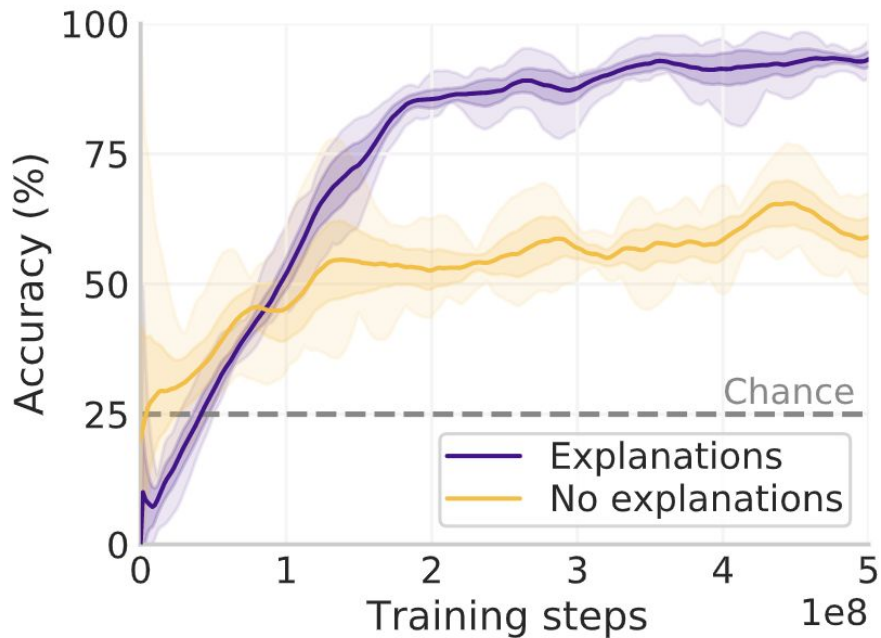
Predicting explanations during training



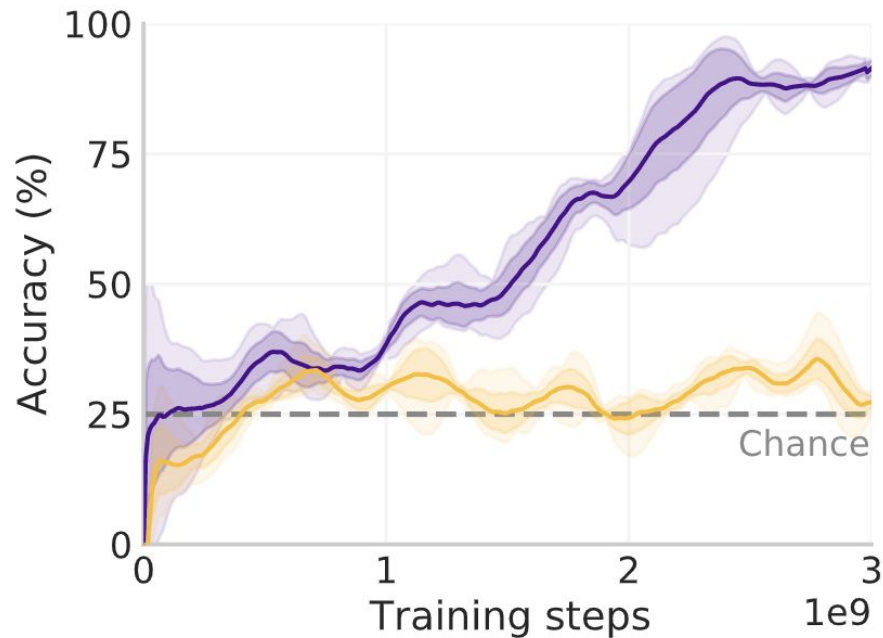
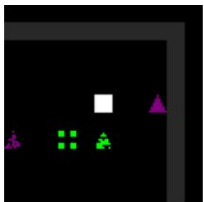
Predicting explanations during training



Explanations improve learning of odd-one-out tasks



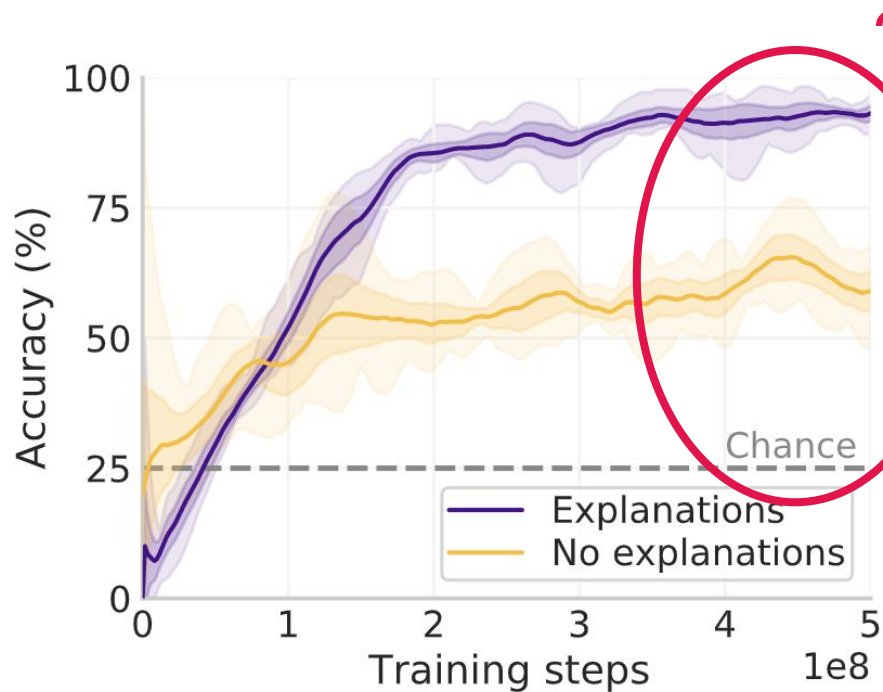
(b) 2D results.



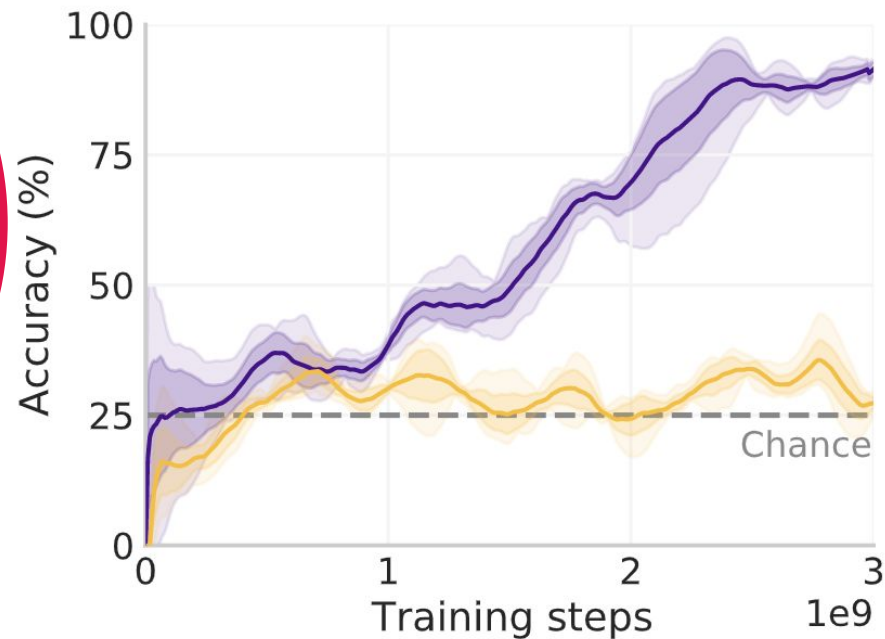
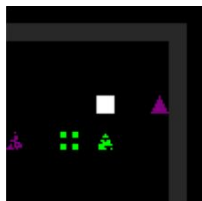
(c) 3D results.



Explanations improve learning of odd-one-out tasks



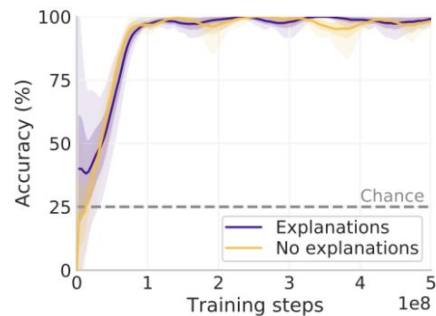
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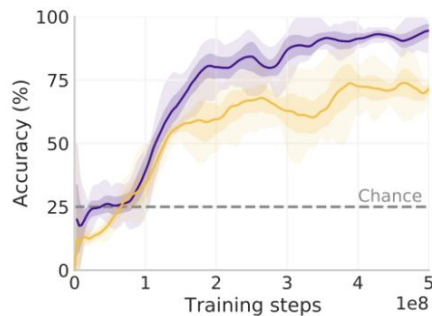
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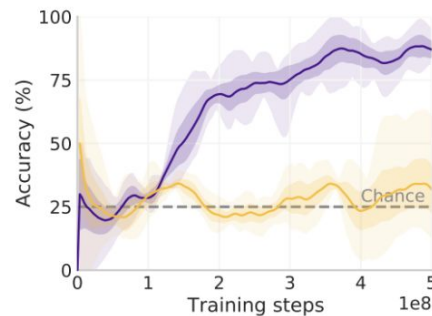
In 2D, explanations help agents move past focus on “shortcut” features to learn more difficult ones



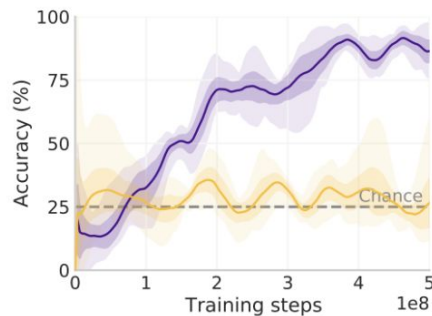
(a) Position.



(b) Color.



(c) Shape.



(d) Texture.

Easier

For CNNs

Harder



Explanations can help with other important challenges (see paper)



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Explanations can overcome
ambiguous training to shape
how an agent generalizes OOD!

Train (confounded):



Evaluation (deconfounded):



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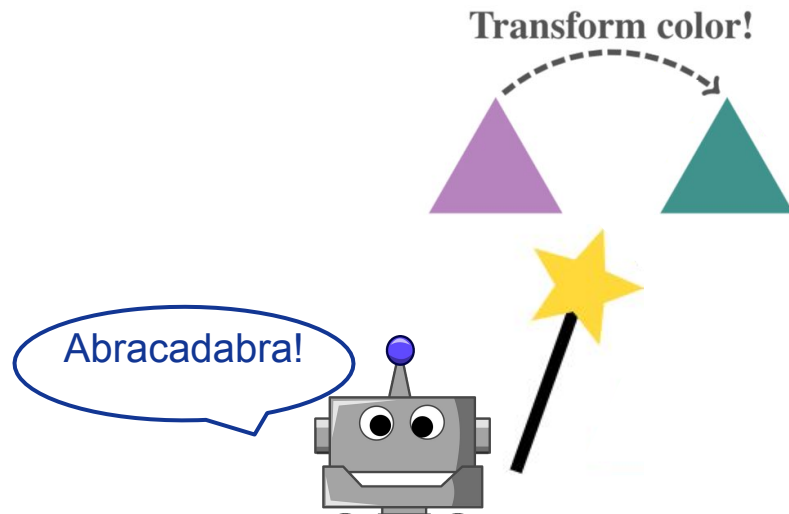
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Explanations allow agents to meta-learn how to perform experiments to identify causality!



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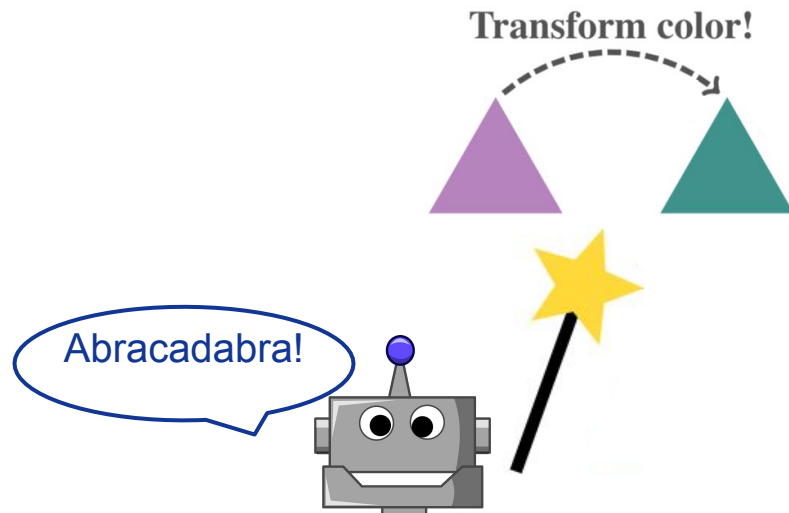
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+ see paper for more analysis, control conditions like unsupervised auxiliary losses, ...

