

# Effects of explaining machine-learned logic programs for human comprehension and discovery

Lun Ai
Department of Computing
Imperial College London

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Imperial College London

**Doctoral researcher** at Imperial College London, UK Interests:

Effects of Al explanations in human-Al collaboration

Sustainable and user-friendly Al to drive science



TAILOR

Supervisor: Stephen Muggleton

Inductive & Abductive Logic Programming

Explainable Al

Computational Scientific Discovery (Biology)

**AI-4-EB Consortium** 

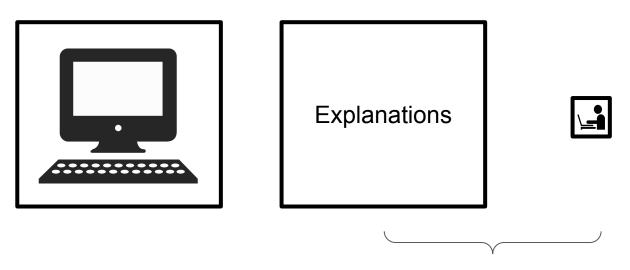
# Explainable AI (XAI) is necessary

[Gunning and Ada, 2019; Miller, 2019; Markus et al. 2021; Minh et al., 2022; Krenn et al. 2022; Schmid and Wrede, 2022; Adamson, 2022]



... for our interactions with AI

## Do users actually understand AI explanations?

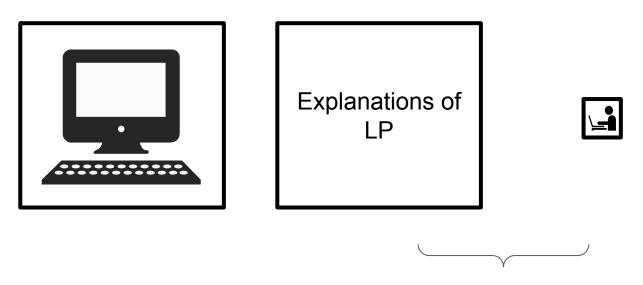


Not quantifiable, e.g. interpretability [Lipton, 2018]

Comprehensibility = model complexity [Guidotti et al., 2018]?

#### "Logic programs are human-understandable"

## Problematic assumption



There are **very few** attempts to understand effects

# Ultra-strong ML -> (beneficial) human behavioural change

Explanatory effect =

machine-aided task performance - self-learning task performance

**Machine-aided:** learning with explanations (e.g. generated from LP)

**Self-learning**: learning with only training examples

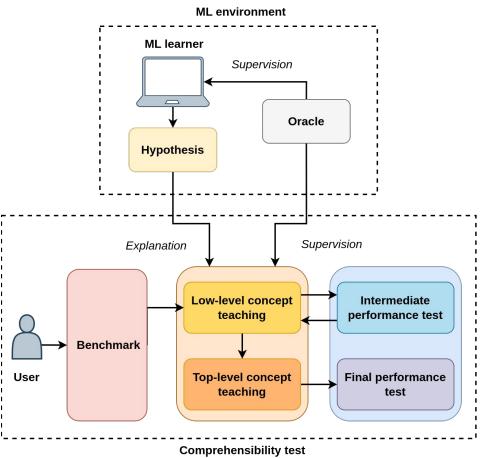
**Performance:** predictive accuracy on unseen test data

## Teaching curriculum

ML: teacher

Human: student

Interactions: curriculum



Comprehensibility test environment

#### Humans are symbol manipulation systems

Cognitive window for a machine-learned logic program P:

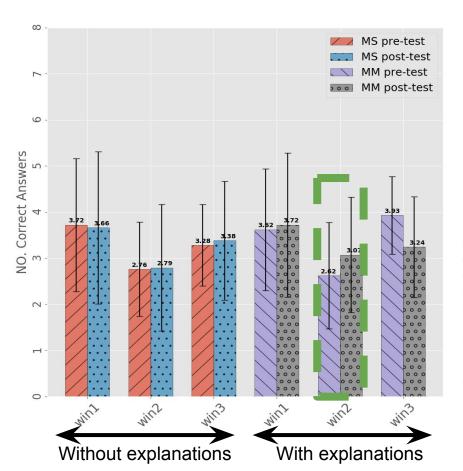
Axiom 1: Hypothesis space to necessarily learn P must be small

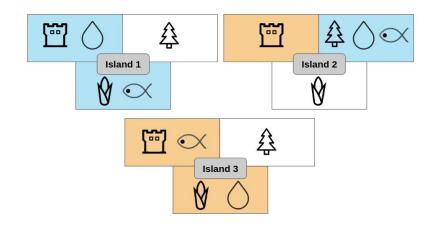
Humans have limited search ability in the hypothesis space

**Axiom 2:** Shortcuts in P to reduce grounding cost (cognitive cost)

Humans have limited capacity for mental computations

#### Cognitive window <u>satisfaction</u> = <u>beneficial</u> effect



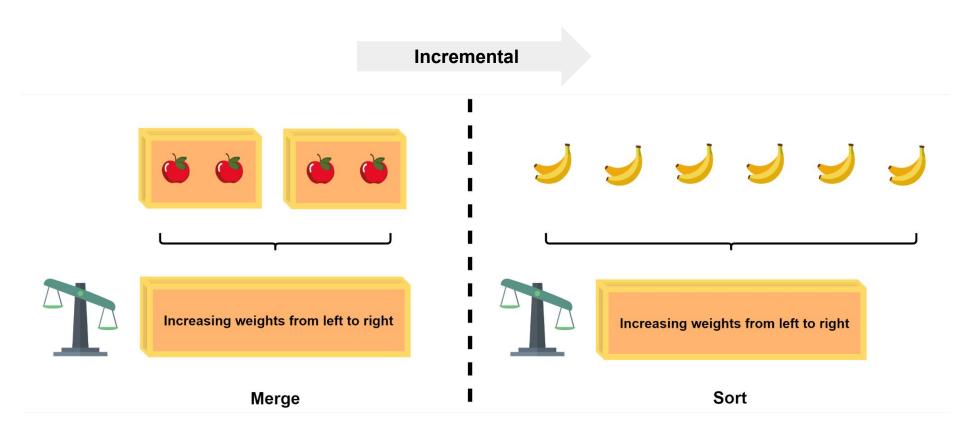


| Depth | Rules  |
|-------|--|
| 1     | $win_1(A,B):-move(A,B),won(B).$  |
| 2     | win_2(A,B):-move(A,B),win_2_1(B).                                      |
|       | <pre>win_2_1(A):-number_of_pairs(A,x,2), number_of_pairs(A,o,0).</pre> |
| 3     | $win_3(A,B):-move(A,B),win_3_1(B).$                                    |
|       | $win_3_1(A):=number_of_pairs(A,x,1),win_3_2(A).$                       |
|       | $win_3_2(A):-move(A,B),win_3_3(B).$                                    |
|       | $win_3_3(A):-number_of_pairs(A,x,0),win_3_4(A).$                       |
|       | $win_3_4(A):-win_2(A,B),win_2_1(B).$                                   |

#### Learned by Metagol

[Ai et al., 2021]

## Teach **Merge Sort** to human novices



#### A variant of **bottom-up** merge sort [Goldstine & Neumann, 1963]

#### Input:

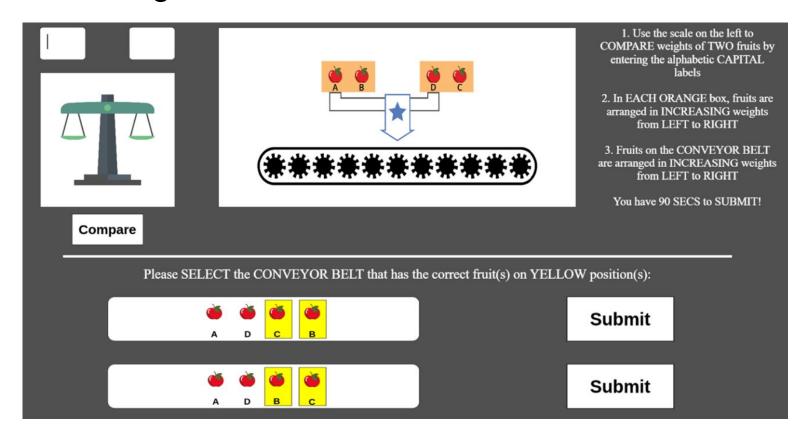
| [4, 6, 5, 2, 3, 1]                                      | Definition                   | Rules   |
|---|------------------------------|---|
| After Iteration 1                                       |                              | (A D) (A C) (A C)   |
| [ <u>4 &lt; 6</u> , <u>2 &lt; 5</u> , <u>1 &lt; 3</u> ] | merger/2                     | <pre>merger(A,B):-parse_exprs(A,C),merger_1(C,B). merger_1(A,B):- compare_nums(A,C),merger_1(C,B) merger_1(A,B):-compare_nums(A,C),drop_bag_remaining(C,B).</pre> |
| After Iteration 2                                       | sorter/2                     | sorter(A,B):-merger(A,C),sorter(C,B).   |
| [2 < 4 < 5 < 6, 1 < 3]                                  | (after learning<br>merger/2) | <pre>sorter(A,B):-recycle_memory(A,C), sorter(C,B). sorter(A,B):-single_expr(A,C), single_expr(C,B).</pre>  |

After Iteration 3

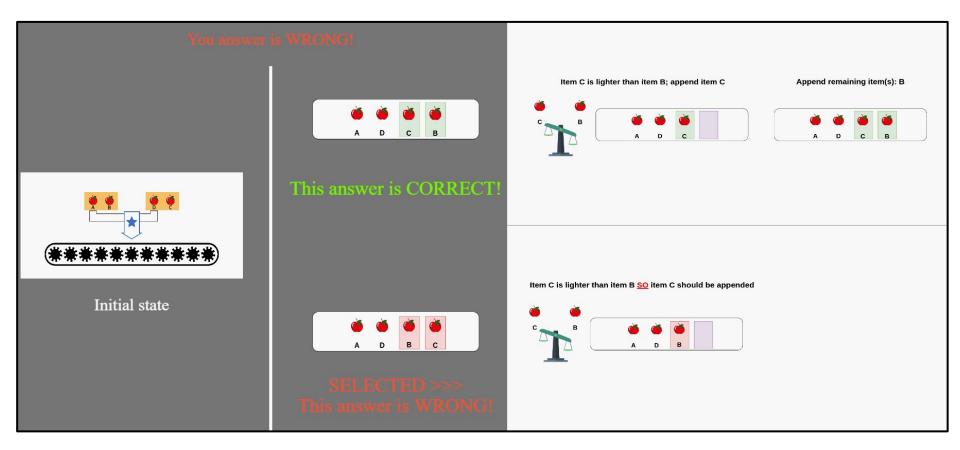
[1 < 2 < 3 < 4 < 5 < 6]

Learned by Metagol

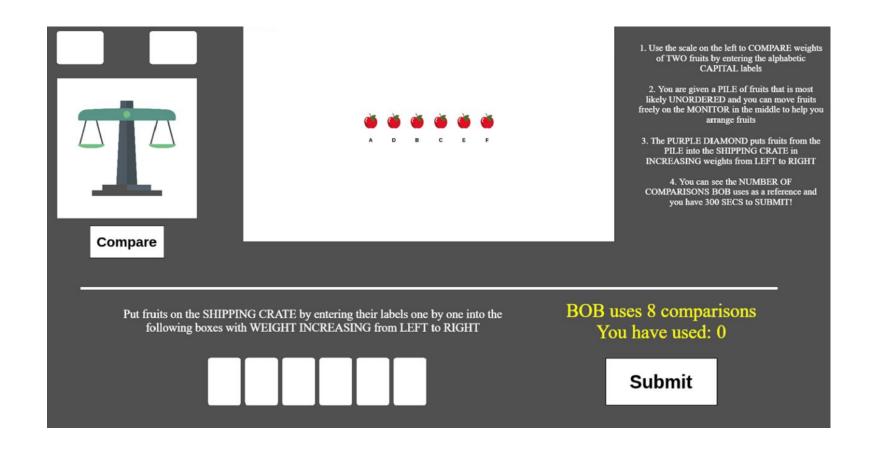
#### Learn to merge



#### Why is/isn't an action optimal?



#### Learn to sort



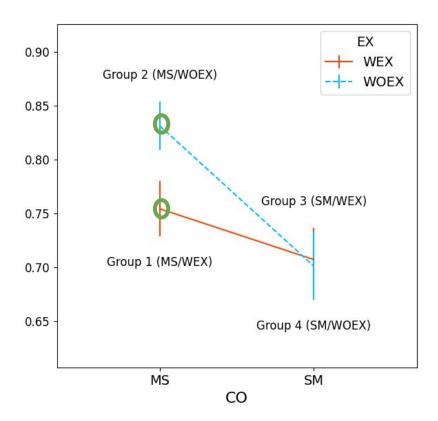
#### Incremental curriculum: improved human performance

PS

Average sorting performance **PS**:

Monotonic correlation of target vs.

human answers (Spearman rank)



# An alternative evaluation? An example.

Sequence [4, 6, 5, 2, 3, 1]

Human trace

$$[(6, 4), (5, 2), (3, 1), (4, 2), (5, 4), (6, 5), (2, 1), (3, 2), (4, 3)]$$

Machine trace (24 algorithms, 6 categories)

$$[(4, 6), (5, 2), (2, 4), (4, 5), (5, 6), (3, 1), (1, 2), (2, 3), (3, 4)]$$

| No. possible comparisons | Not in human trace | In human trace |  |
|--------------------------|--------------------|----------------|--|
| Not in machine trace     | 13                 | 1              |  |
| In machine trace         | 1                  | 10             |  |

 $\chi^2$  = 14.3 with p < .001 and Spearman rank correlation  $\rho$  =.9 and p < .001

# Novel strategy adaptation: quick sort

| PS                     | BS   | DS   | IS   | MS   | QS   | Hybrid | Other |
|------------------------|------|------|------|------|------|--------|-------|
| $Group \ 1 \ (MS/WEX)$ | _    | _    | _    | _    | _    | _      | _     |
| Training               | .012 | .075 | .150 | .000 | .175 | .162   | .425  |
| Performance test       | .056 | .094 | .162 | .025 | .238 | .175   | .250  |
| Differences            | .044 | .019 | .012 | .025 | .063 | .013   | 175   |
|                        |      |      |      |      |      |        |       |

Unexpected efficient strategy with **better** performance (incremental learning with explanations)

#### Messages & Open questions

- 1. LPs are not always human-comprehensible
  - How do we optimise comprehensibility
  - Is possible to formulate a theory of incomprehensibility?
- 2. We can learn a lot by studying effects of LP explanations
  - What insights can we get from human trace and ILP learner trace?
  - How can we design curricula to enable human discovery?
- 3. There are **limitations** to performance-based evaluations
  - How should we evaluate strategy adaptations?

Lun Ai

Email: <u>lun.ai15@imperial.ac.uk</u>

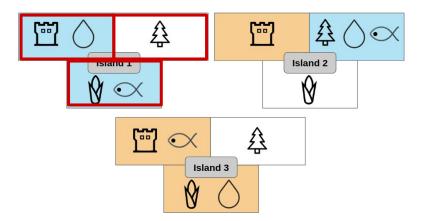
Linkedin: https://www.linkedin.com/in/lun-ai-46481a128/

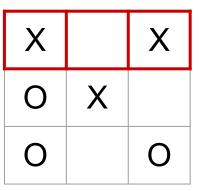
Website: <a href="https://lai1997.github.io/">https://lai1997.github.io/</a>

## Isomorphism of Noughts and Crosses

You play Blue, and please press a **WHITE** cell to capture resources that you think can lead to WIN You have **ONE CHANCE** for each question.

#### Question NO.1

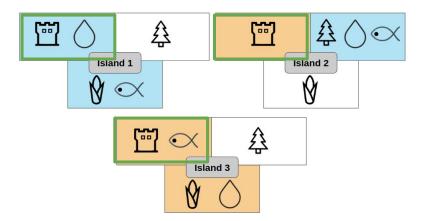


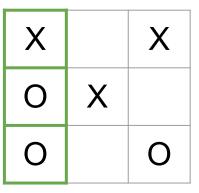


## Isomorphism of Noughts and Crosses

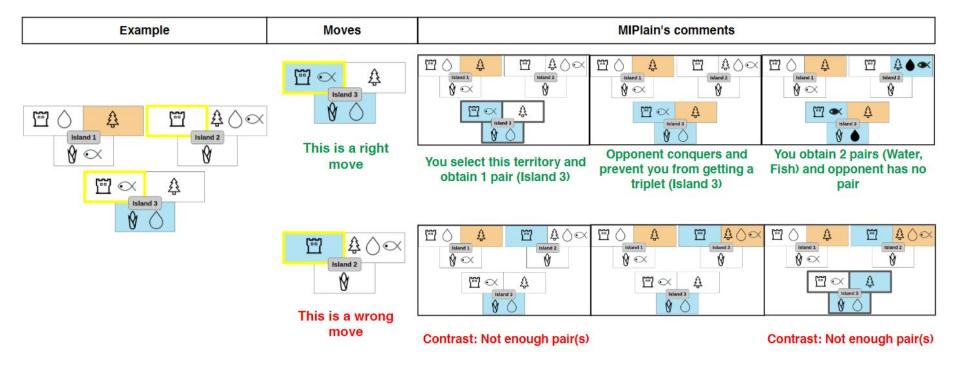
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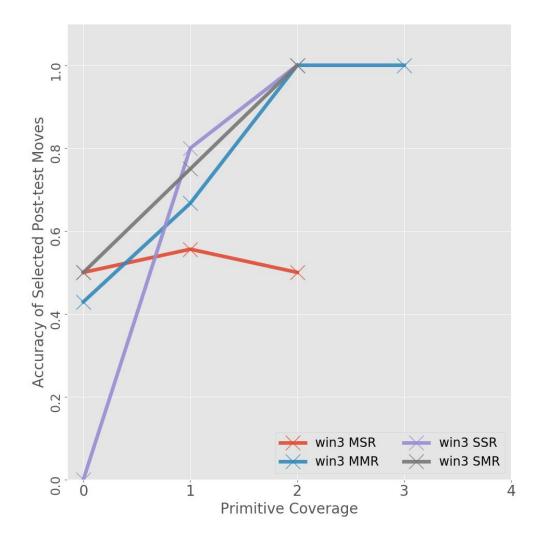
#### Question NO.1





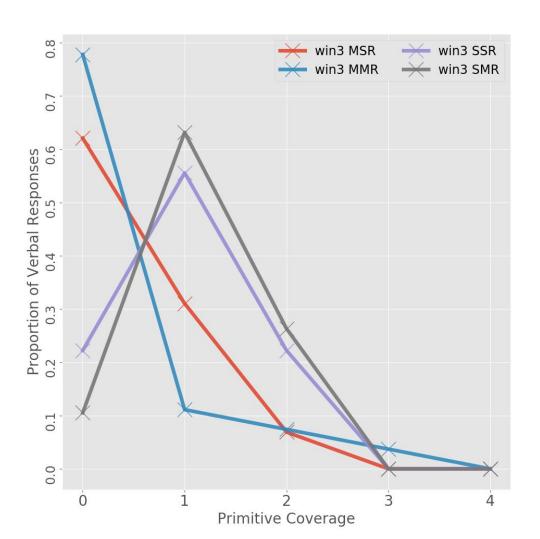
#### Learn the Island Game (Noughts and Crosses isomorphism)





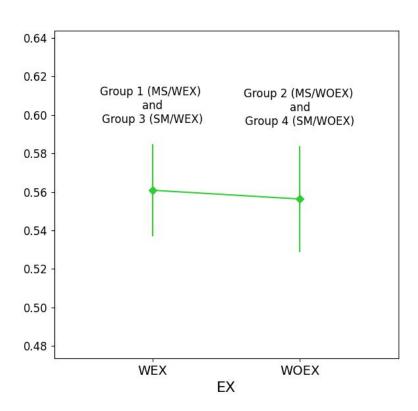
**Primitive coverage: No.**descriptions of primitives in textual responses

High correlation with performance



Low frequency of high coverage (key predicates) responses

# No performance change by explaining merge



## Curriculum arrangements

