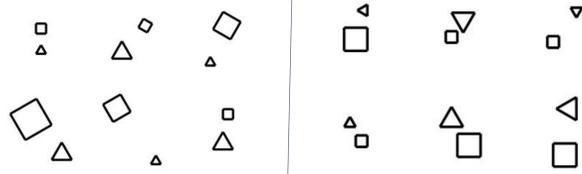


Bongard Problems

Visual Reasoning tasks with 6 positive and 6 negative example images for a particular concept. Given the examples images, the task is to find the differentiating concept. For example -

Negative Examples

Positive Examples



Concept: **Triangles above squares.**

We evaluate system on adaptations of: #4, #14, #16, #21, #23, #24, #36, #40, #53, #60, #75, #85, #94 and #96 from www.foundalis.com/res/bps

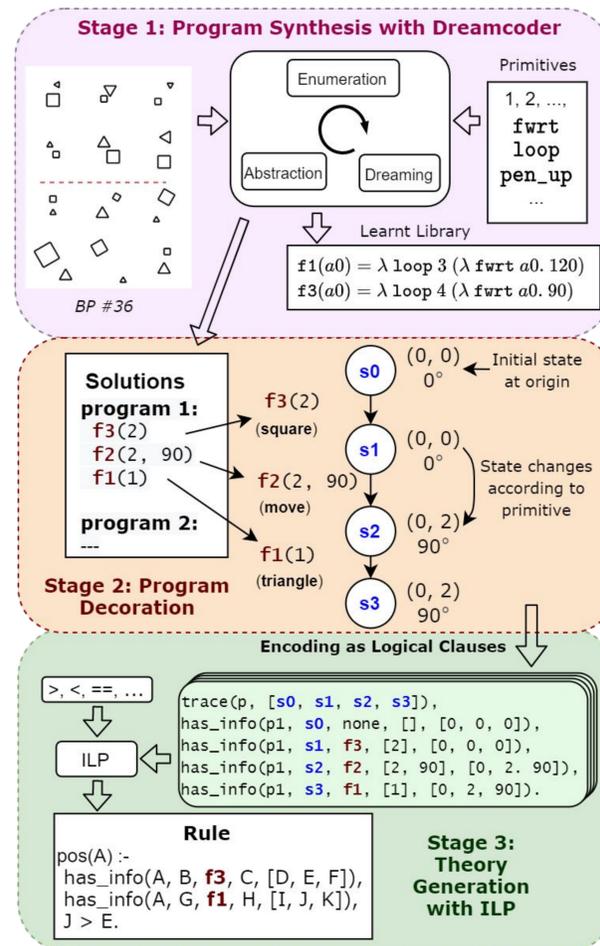
Overview

Solving reasoning tasks requires the use of suitable representations which can encapsulate relevant concepts. Such a representation should also allow flexibility in abstraction formation at various levels in the hierarchy.

In our 3-staged inductive programming system, we use decorated graphical programs to represent the images for Bongard Problems. We postulate that this allows for formation of concepts:

- At the first stage through invention by abstraction in functional λ -calculus programs using Dreamcoder. Such as learning a polygon from line instructions.
- By allowing for additional methods of information extraction from the solution program at the decoration phase during a debugger style step-by-step execution of the λ -calculus programs.
- On top of the decorated functional λ -calculus programs, using logical programs, through Inductive Logic Programming, to learn higher level concepts such as *Triangle above Square*, *Concavity / Convexity*

Method



1. **Synthesis:** All positive and negative images for a Bongard Problem are input to Dreamcoder to obtain a library of higher level primitives and solution programs for each image
2. **Decoration:** Each Program is converted into a decorated state transition diagram via a *debugger*-styled execution. Transitions are decorated with primitive calls and states with information such as the position and orientation of the cursor during the program execution.
3. **Theory Identification:** A FOL representation of the transitions using `has_info` and `trace` predicates, along with comparison predicates, are input, as background knowledge, to Aleph (an ILP Engine) to find a theory that differentiates between the positives and negatives.

Predicate Definitions

has_info(+Program, -State, #Primitive, -Args, [-X, -Y, -Angle])
trace(+Program, [-state0, -state1, -state2, ...])

Results

Our system is able to solve 8 of the 14 Bongard Problems considered. Some are illustrated below -

Concept	Invented Primitives (Dreamcoder)	Theory	Explanation
Anti-clockwise vs Clockwise BP #16	$f2(a0), f3(a0)$: both draw anti-clockwise spirals with different step lengths and with $a0$ controlling tightness of the spiral.	$\text{pos}(A) :- \text{has_info}(A, B, f3, C, [D, E, F]).$ $\text{pos}(A) :- \text{has_info}(A, B, f2, C, [D, E, F]).$	Presence of invented primitive for drawing spirals that are anticlockwise.
Smaller shape present BP #21	$f1(a0, a1)$: Draw an $a0$ -sided polygon with sides of length $a1$	$\text{pos}(A) :- \text{has_info}(A, B, \text{rtfwdt}, C, [D, E, F]), C = [G H], H = [I J], G > I, \text{has_info}(A, K, f1, L, [D, E, F]).$ $\text{pos}(A) :- \text{has_info}(A, B, f1, C, [D, E, F]), C = [G H], H = [I J], G > I.$	Program contains a move primitive where the division factor for angle is greater than multiplication factor for distance, or there is a polygon with side length less than number of sides. Indicating the shape is small.
Triangle above Square BP #36	$f1(a0)$: Draws triangle of side length $a0$. $f3(a0)$: Draws square of side length $a0$	$\text{pos}(A) :- \text{has_info}(A, B, f3, C, [D, E, F]), \text{has_info}(A, G, f1, H, [I, J, K]), J > E.$	Triangle exists with y coordinate greater than that of square
Enclosed shape has fewer sides BP #53	$f1(a0, a1)$: Draw an $a0$ -sided polygon with sides of length $a1$	$\text{pos}(A) :- \text{has_info}(A, B, f1, C, [D, E, F]), \text{has_info}(A, R, \text{pt}, Q, [K, L, M]), \text{has_info}(A, I, f1, J, [K, L, M]), C = [G H], J = [N O], O = [P Q], G > N, N > P.$	Smaller polygon (having length of side smaller than number of sides) has fewer sides than larger (enclosing) polygon.

The main reasons where the system fails are -

- **Representation:** Inability to represent solid fills, arbitrary curves and other irregular features using current DSL.
- **Search:** High number of shapes / lines to be drawn meaning intractable search due to large program lengths of the solution.

Next Steps

Our work can be improved in 3 key areas:

- **Graphical Program Synthesis:**
 - A learned metric for comparison rather than pixel level comparison
 - Execution-guided synthesis rather than enumeration
- **State Decorations:**
 - Learned automated feature extractors to work on top of produced programs/images.
- **Final theory learning step:**
 - Construction of meta-rules among programs of different problems to learn general concepts such as *smallness*, etc
 - Construction of meta-rules, in the 2nd order, over sub-programs of the same problem