

Using Imagery to Simplify Perception

29th Soar Workshop

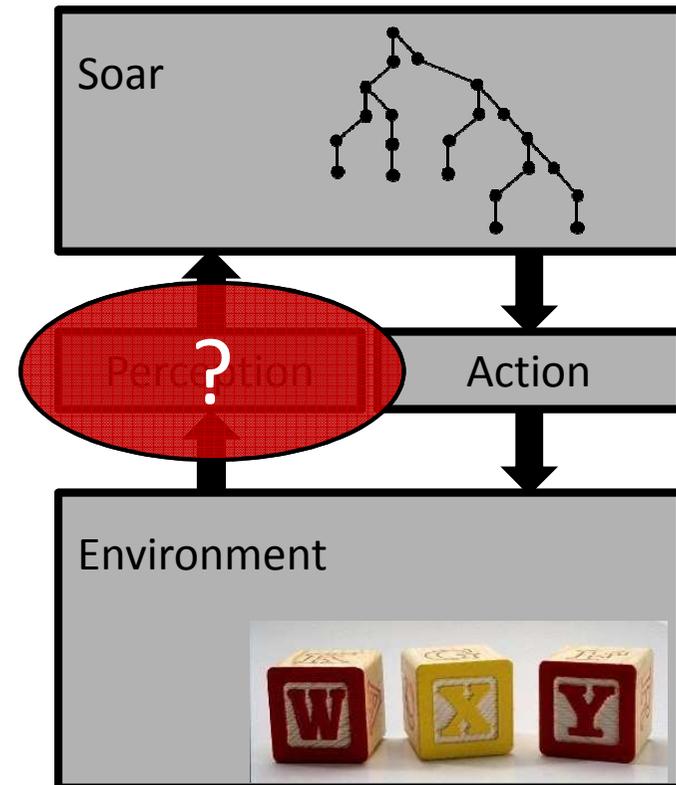
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SVS Project Background

- ▶ Soar Visual/Spatial (SVS) adds visual and spatial processing to Soar
- ▶ Previous workshops:
 - ▶ Architecture was presented
 - ▶ Pieces of it were examined
 - ▶ Agents were shown
- ▶ This year: studying *sufficiency* and *generality* in representing spatial problems
 - ▶ Giving Soar the tools it needs to solve lots of spatial problems

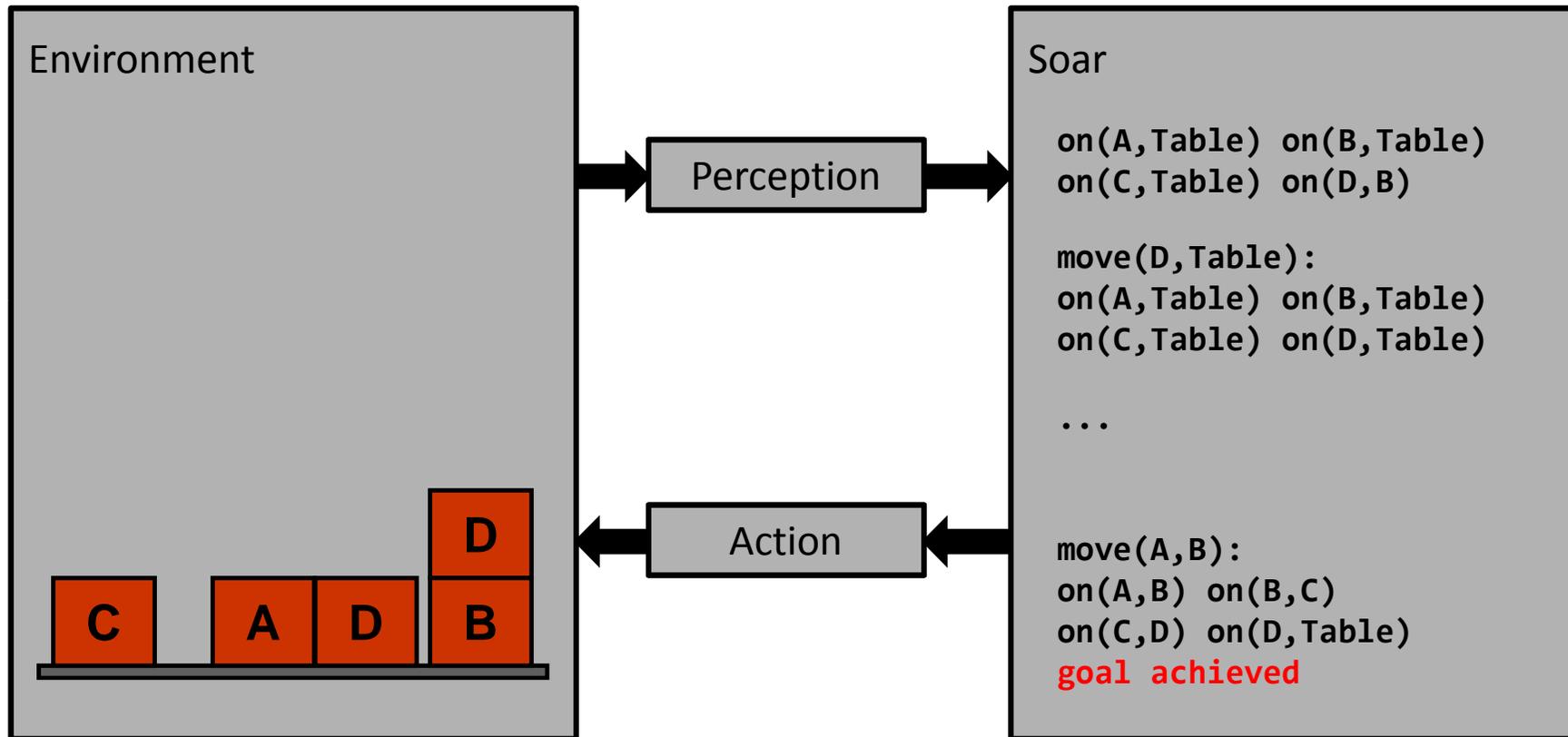
Getting Symbols from Sensors

- ▶ Goal: Allow Soar to solve arbitrary spatial problems
- ▶ Problem state information in Soar is (almost always) abstract
- ▶ Basic perceptions available to an embodied agent might have much more detail
- ▶ What perceptual information should be provided to Soar?
- ▶ This problem is about *what* is calculated by the perception system, not about *how* to calculate it



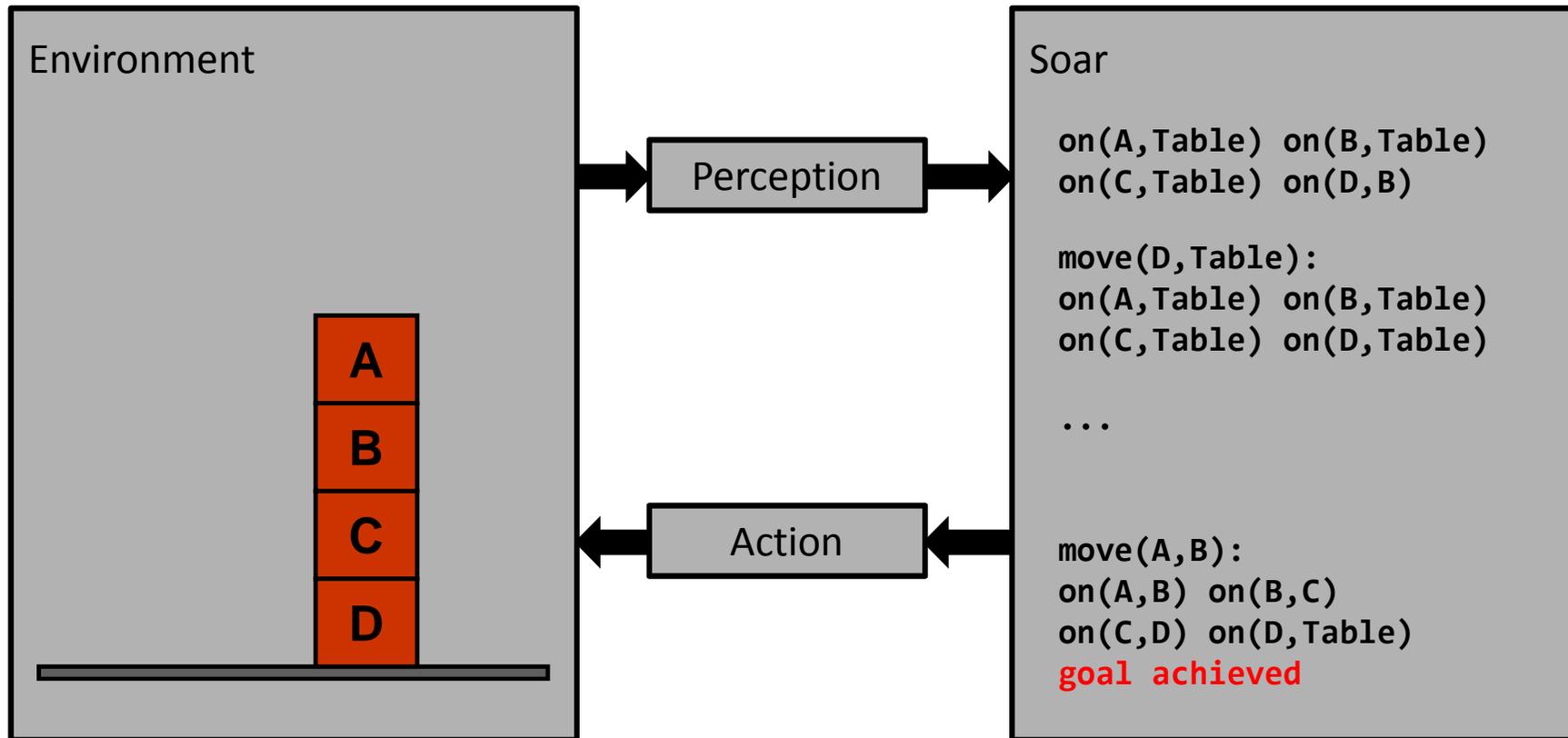
Motivating Example (1)

- ▶ Symbolic Planning in the Blocks World



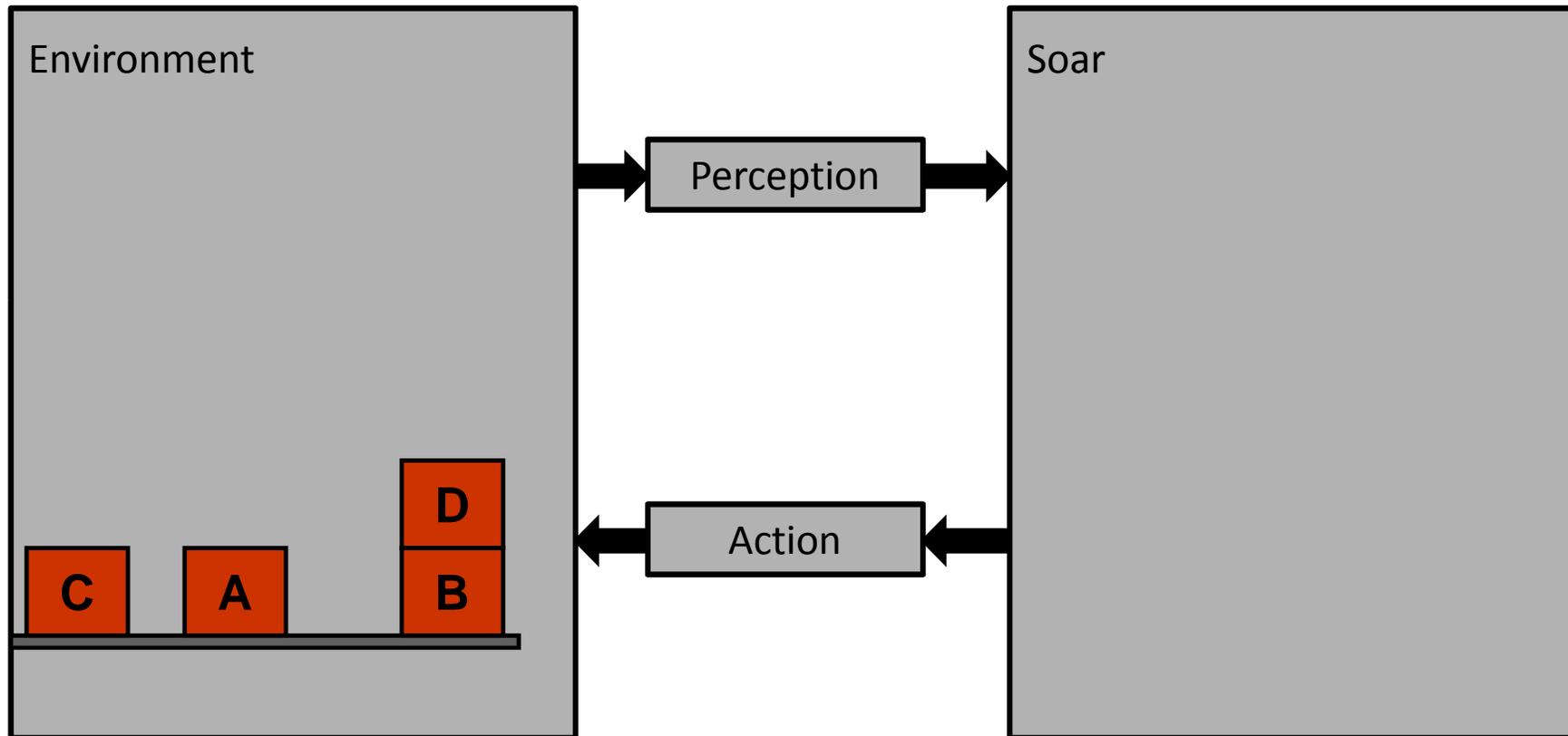
Motivating Example (1)

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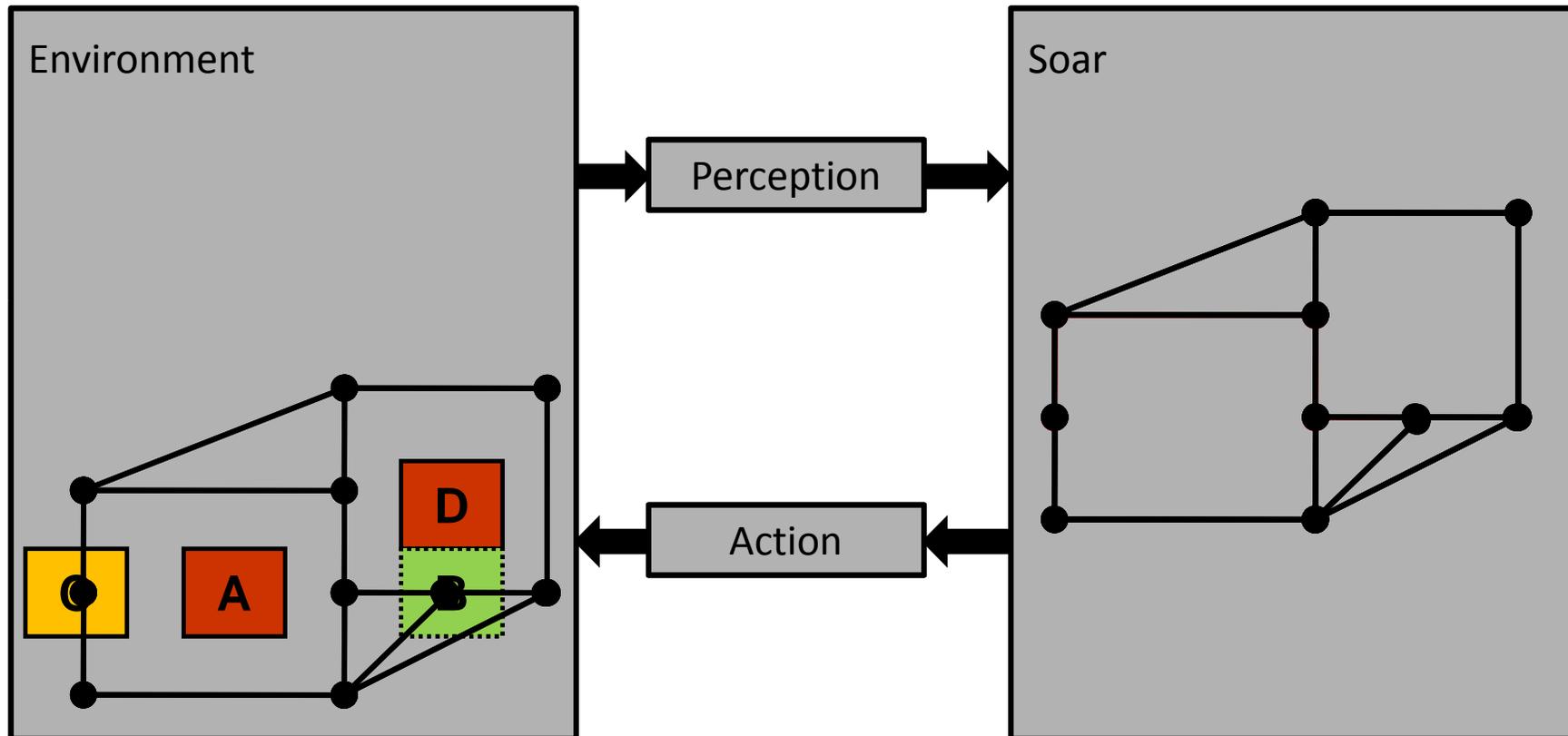
Motivating Example (2)

- ▶ Motion Planning



Motivating Example (2)

- ▶ Motion Planning (**robot**, **obstacles**, **goal**)

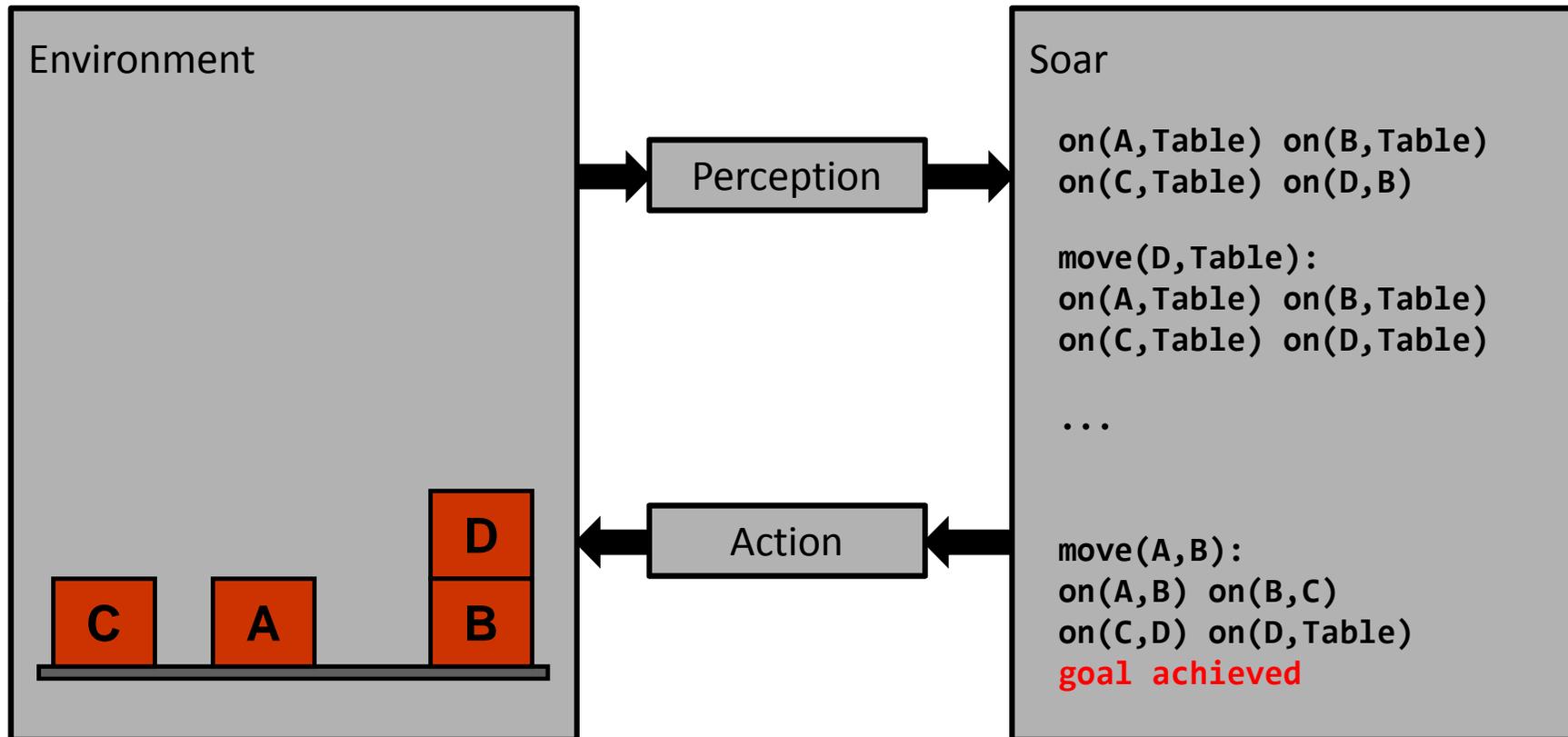


Problem Dependencies in Perception

- ▶ Problems can be geometrically similar, but require very different symbolic abstractions
 - ▶ Any given agent only has one perception system
 - ▶ An agent may need to solve unforeseen problems
- ▶ Possible solution: space -> symbol transformation can be considered task knowledge.
 - ▶ Soar rules for blocks world would map a set of coordinates to “on” relationships
 - ▶ Soar rules for navigation would take a set of coordinates, calculate configuration space, and determine locations and adjacencies
 - ▶ This is certainly possible, but math-intensive and hard to learn
- ▶ Possible solution: abstractions can be symbolically composed from problem-independent primitives
- ▶ These domains are also very simple...

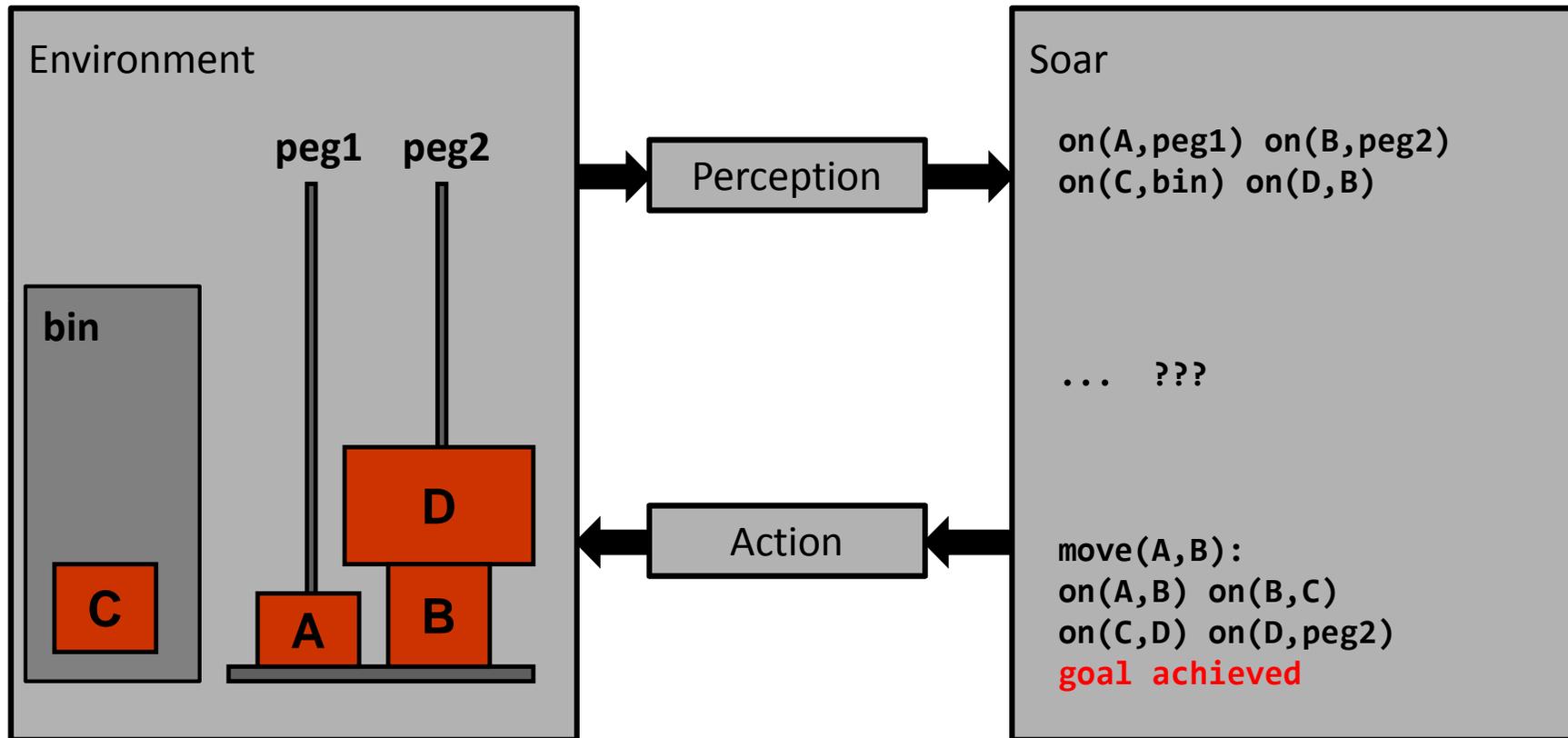
Motivating Example (3)

- ▶ Symbolic Planning in the Blocks World



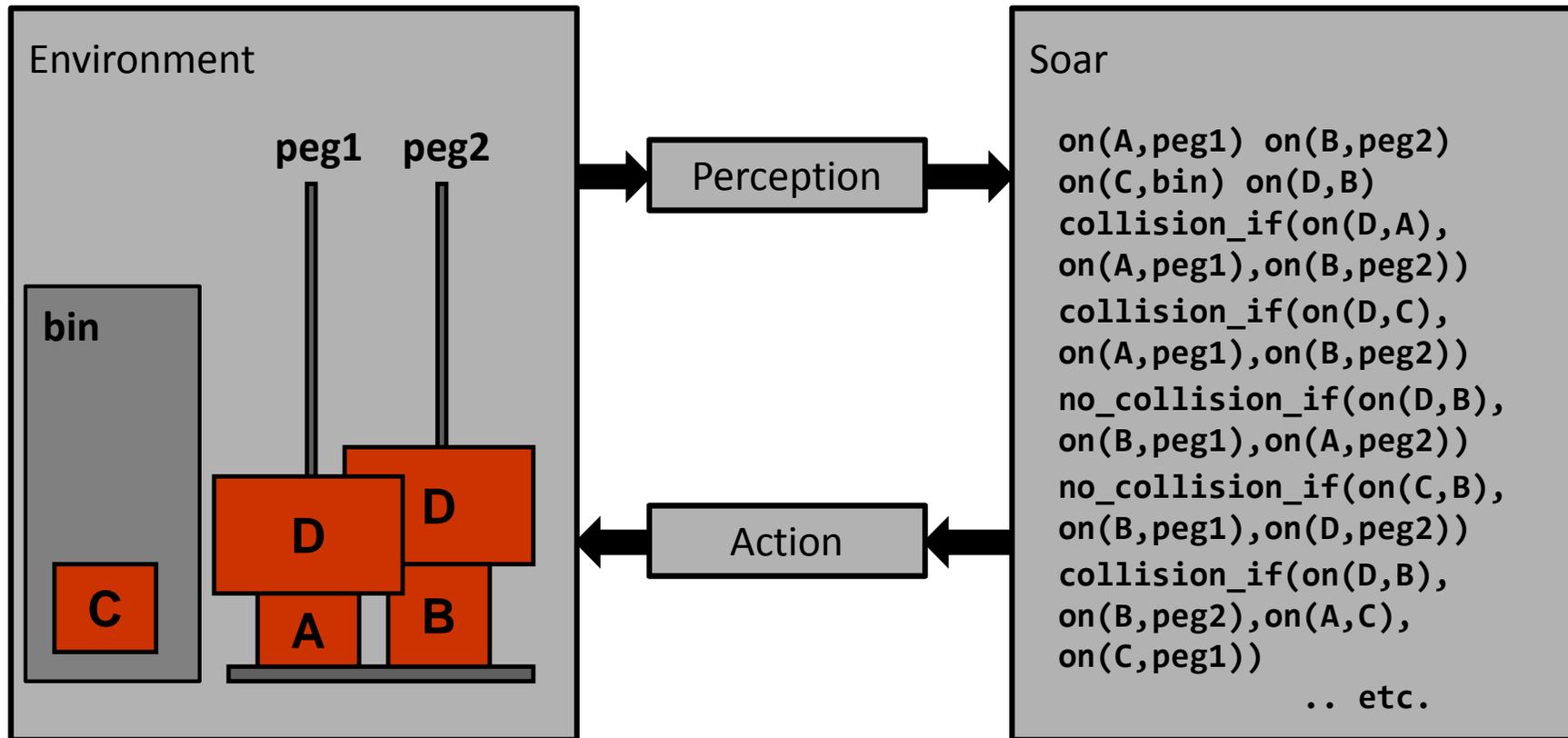
Motivating Example (3)

- ▶ Symbolic Planning in the Pegged Blocks World



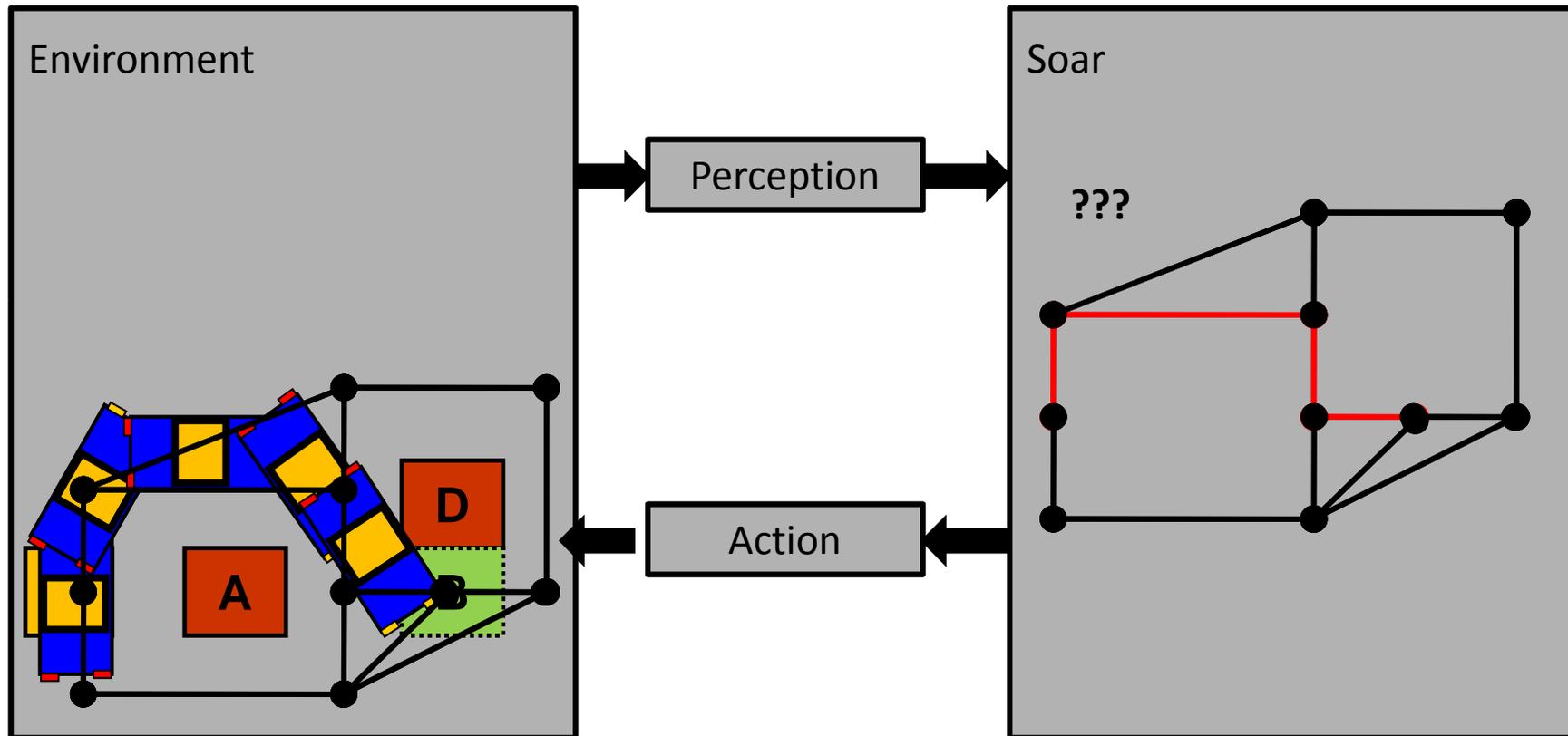
Motivating Example (3)

- ▶ Symbolic Planning in the Pegged Blocks World



Motivating Example (4)

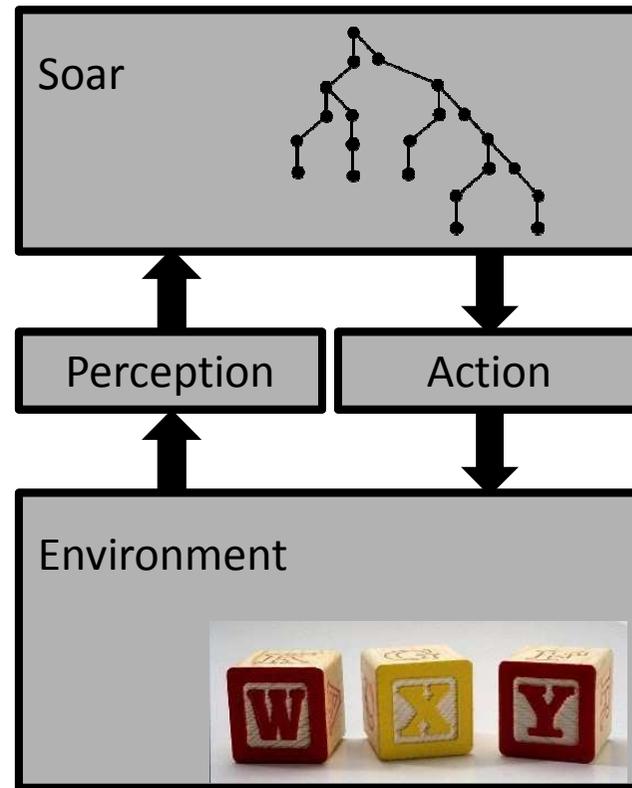
- ▶ Non-Holonomic Motion Planning (**robot**, **obstacles**, **goal**)



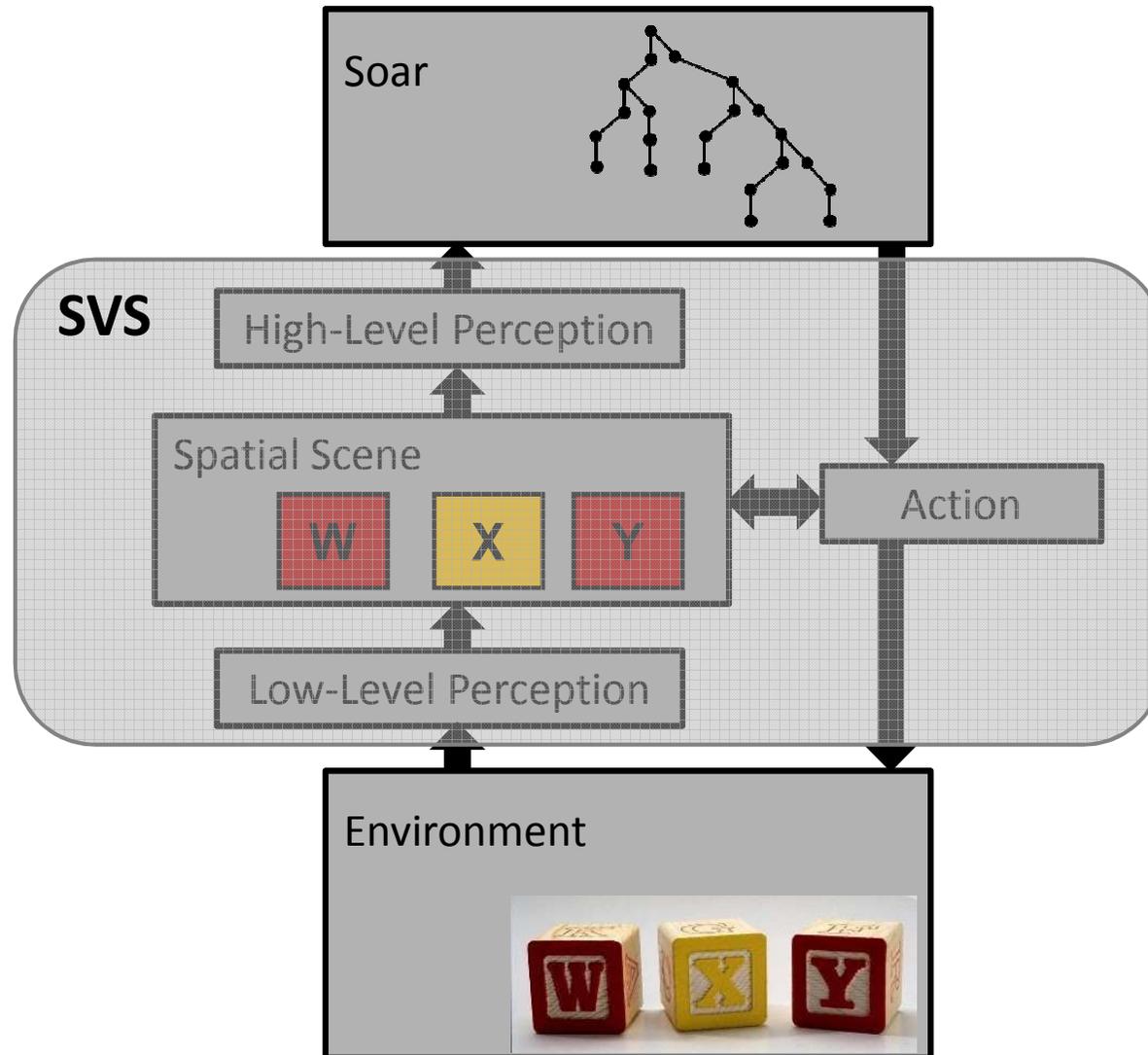
Difficulties in perception

- ▶ Different problem domains require very different qualitative symbolic relationships
 - ▶ But any given agent only has one perception system
- ▶ For some problems, computing *any* useful qualitative symbolic representation is difficult (or impossible)
 - ▶ And if it is possible, the calculated relationships will be extremely problem-specific
- ▶ Solution: reconsider one-way perception

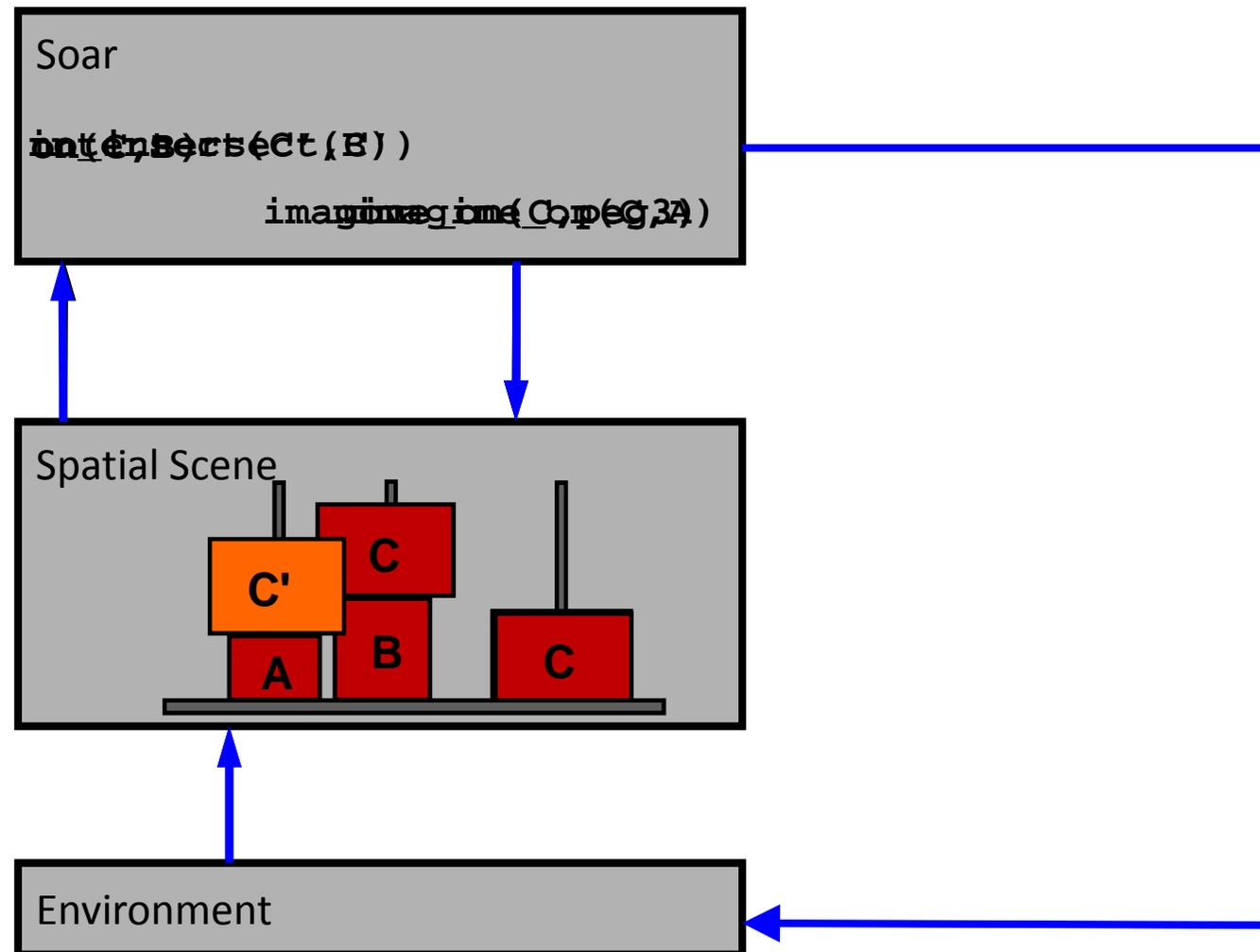
Imagery



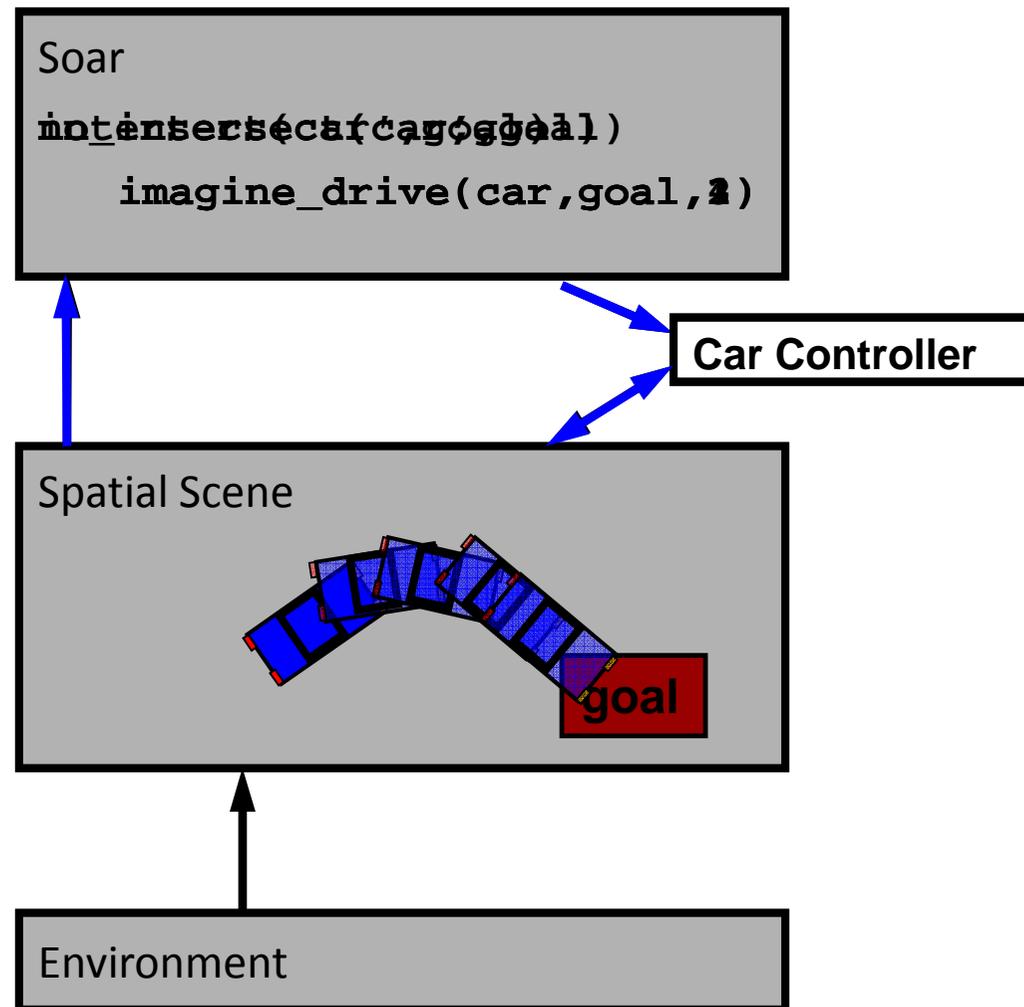
Imagery



Imagery in the Pegged Blocks World



Imagery for Non-Holonomic Motion

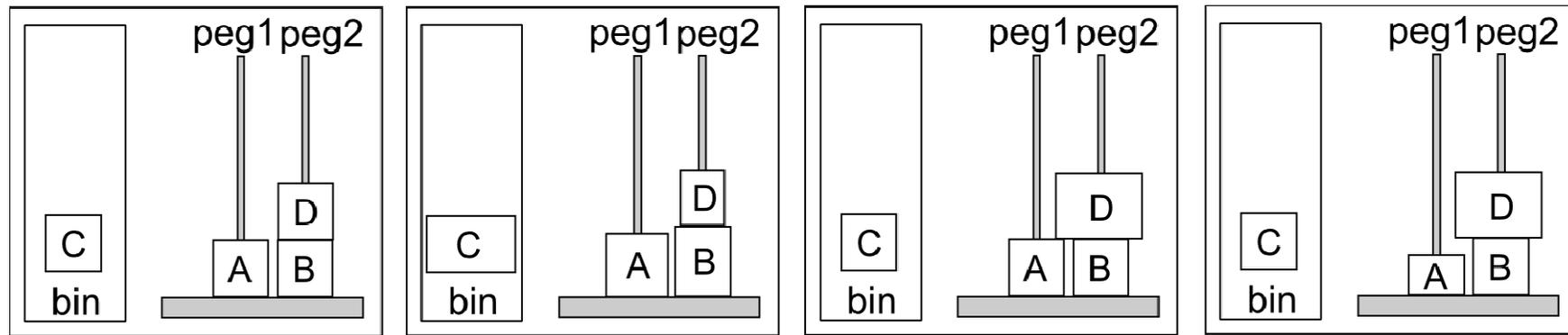


Soar / Imagery Interface

- ▶ High-level perceptions to Soar:
 - ▶ Object identities
 - ▶ Object topology (intersecting or not)
 - ▶ Object distances
 - ▶ Object directions (left-of, right-of, etc.)
- ▶ Imagery actions:
 - ▶ Qualitative predicate projection (e.g., imagine A on B)
 - ▶ Motion simulation
 - ▶ Memory retrieval

Pegged Blocks World Agent

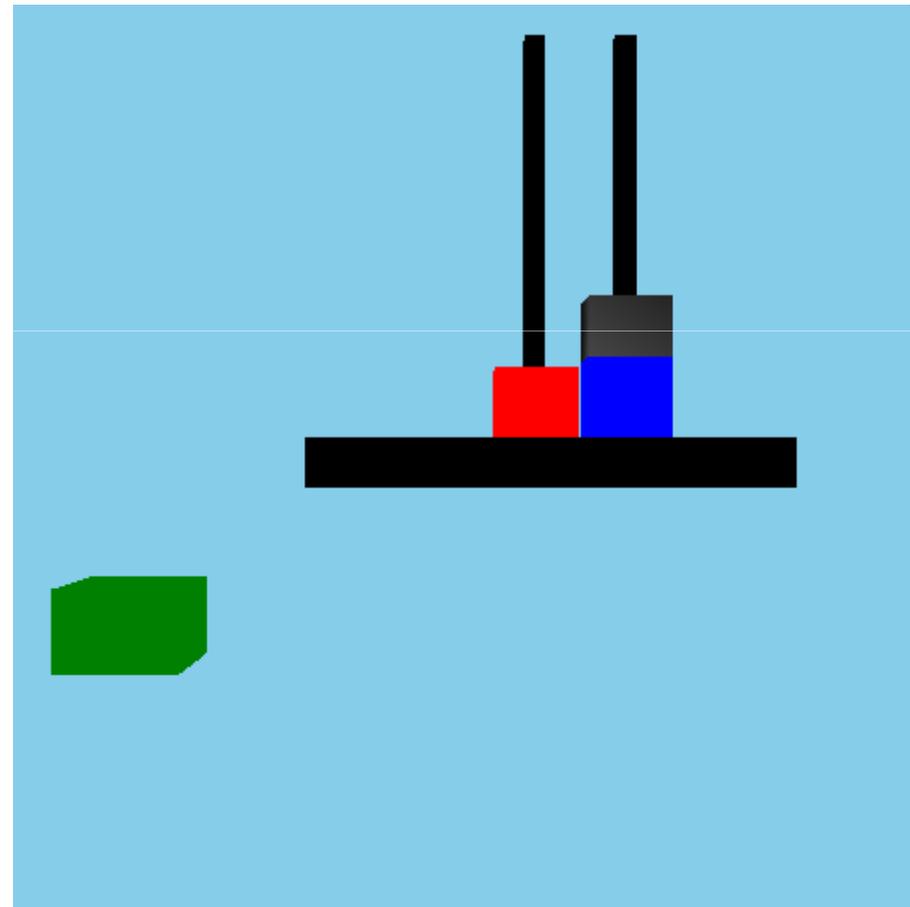
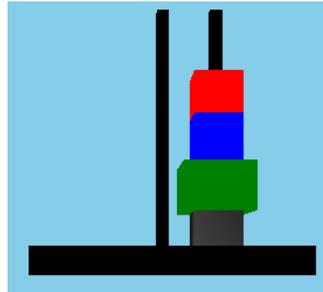
- ▶ All instances considered have the same abstract initial state and goal
- ▶ Any instance may be in one of four cases where the optimal plan differs



- ▶ Soar with SVS can encode a plan to get (almost) optimal behavior in all four cases
 - ▶ This would be extremely difficult without imagery

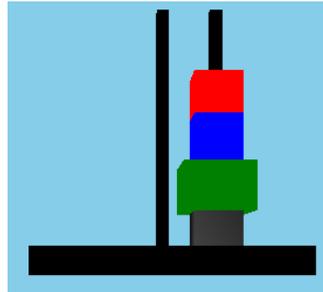
Pegged Blocks World Agent

▶ goal :

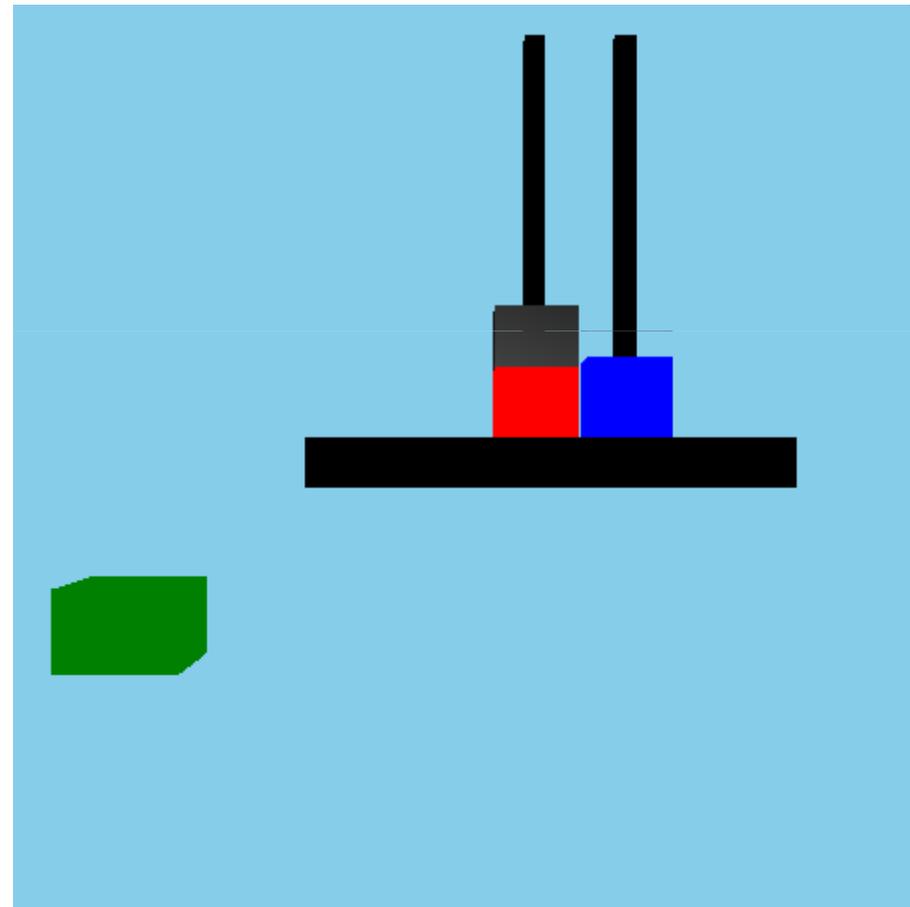


Pegged Blocks World Agent

- ▶ goal :

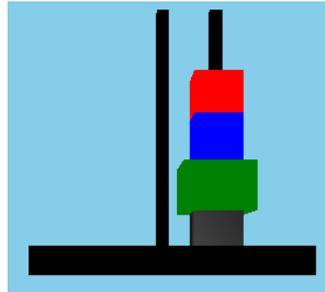


- ▶ move black above red, if it fits

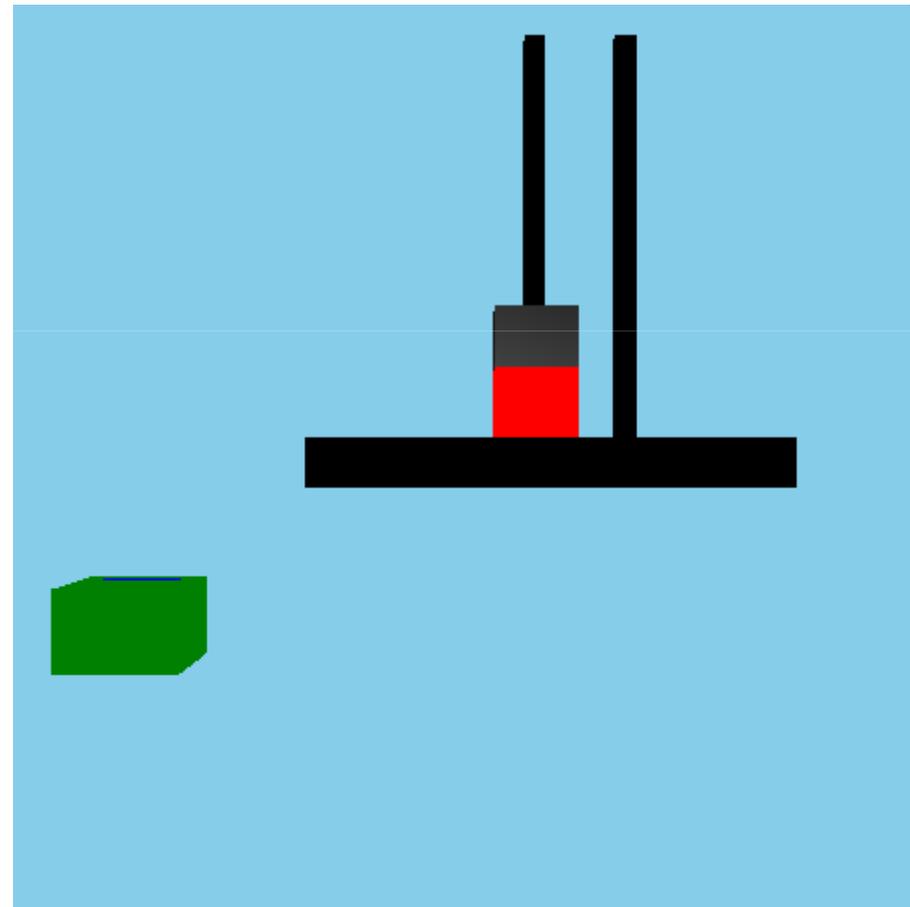


Pegged Blocks World Agent

▶ goal :

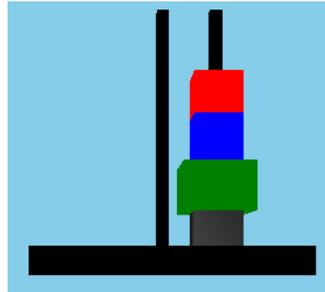


▶ move blue to bin

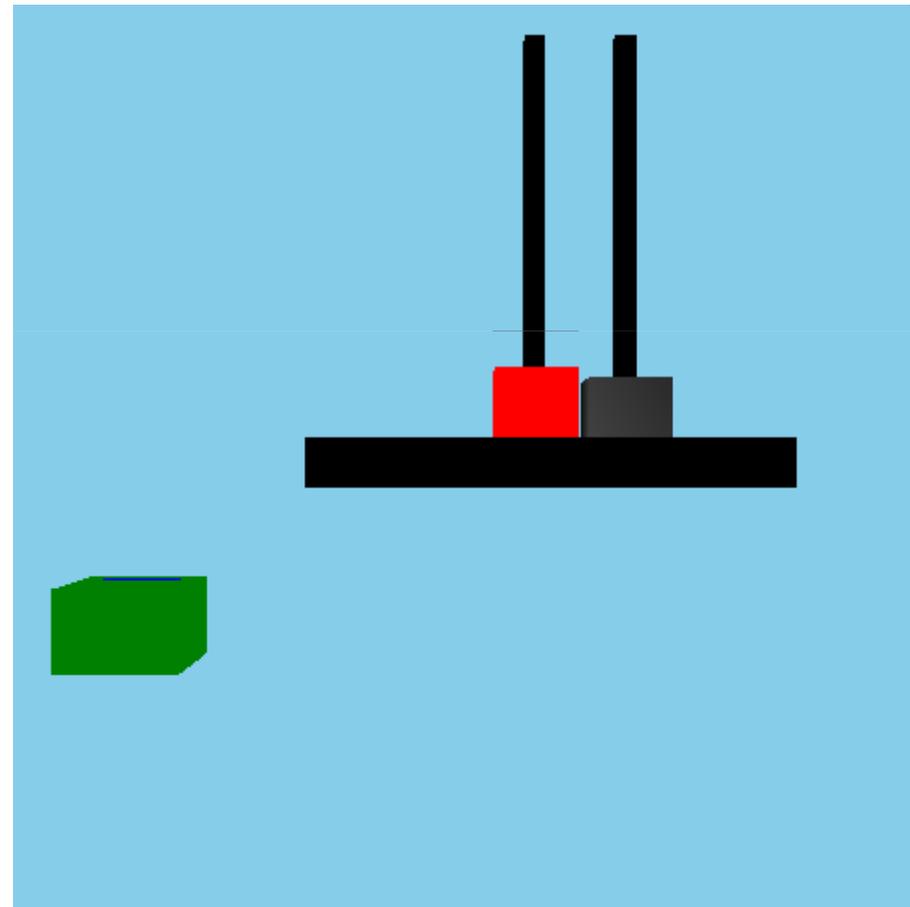


Pegged Blocks World Agent

- ▶ goal :

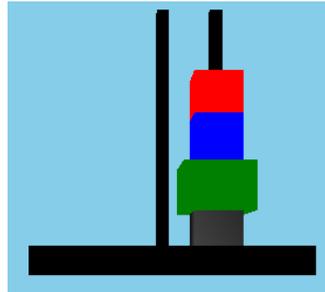


- ▶ move black next to red, if it fits

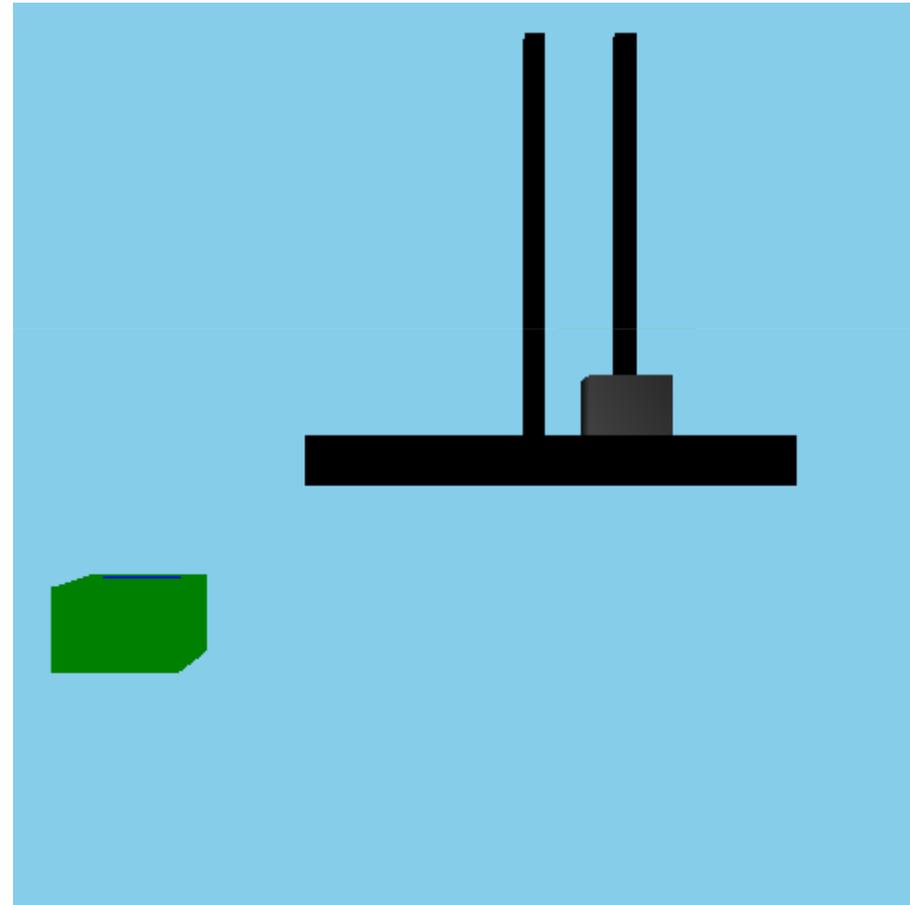


Pegged Blocks World Agent

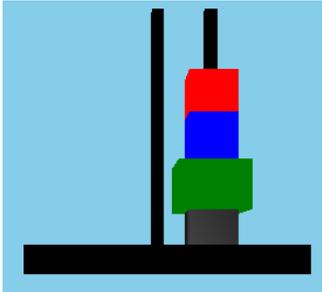
- ▶ goal :

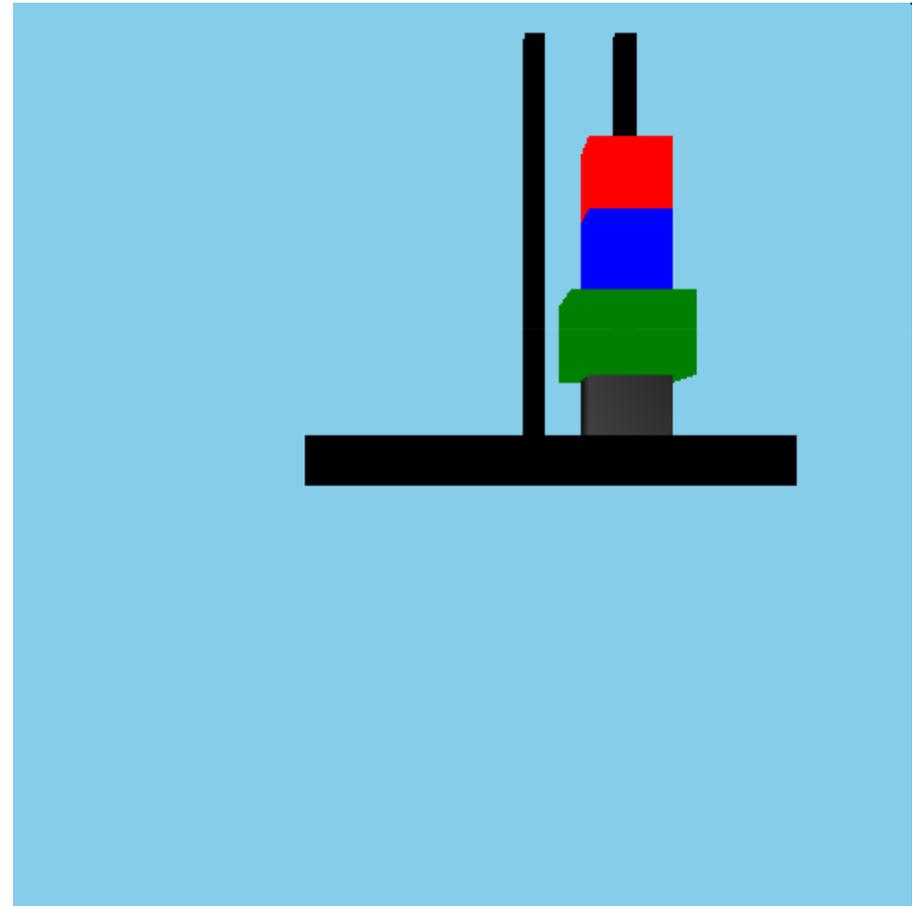


- ▶ move green above black, if it fits
- ▶ else, move red to bin



Pegged Blocks World Agent

- ▶ goal : 
- ▶ build goal stack from bin

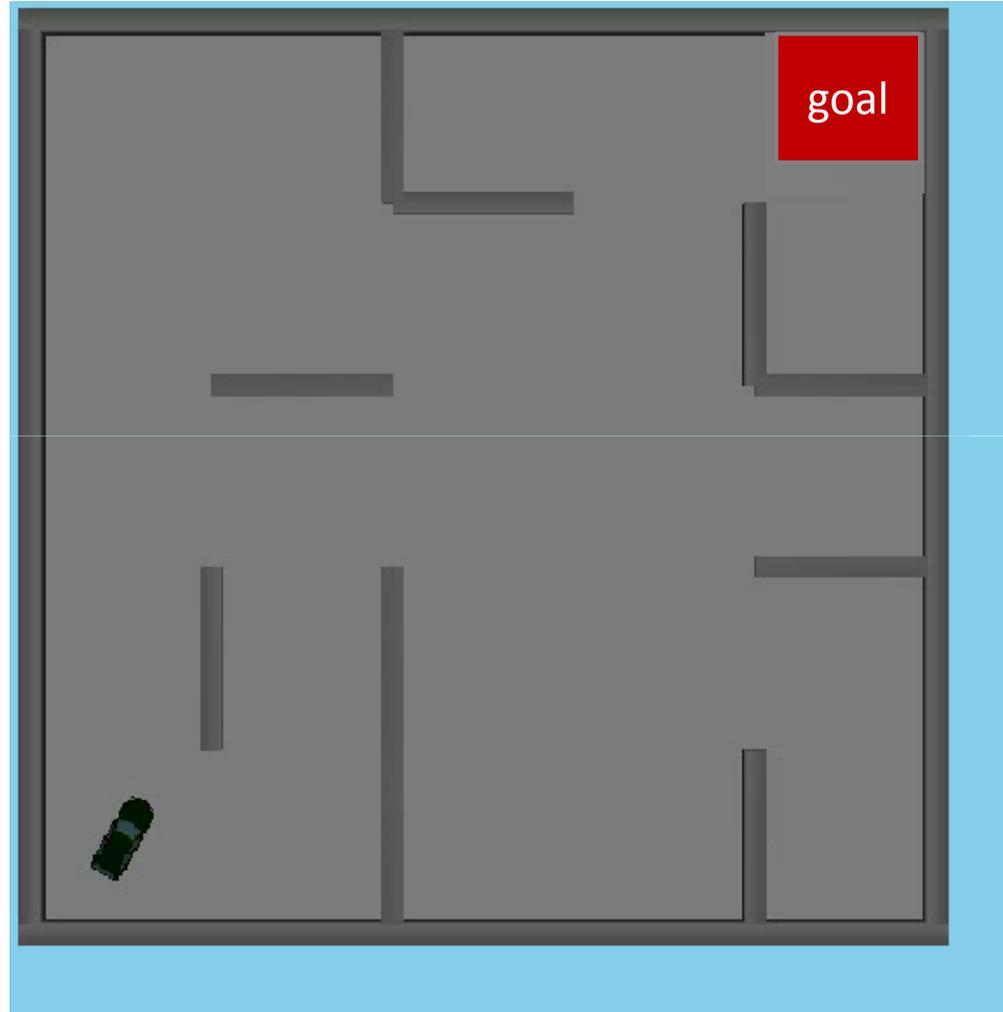


Motion Planning Agent

- ▶ Common robotics problem: must determine a sequence of actions to move from place to place in a fully-observed world
- ▶ Abstraction can be very difficult
- ▶ Result: sampling-based motion planning
 - ▶ RRT Algorithm: through simulation, build a tree of reachable configurations until the goal is reached
- ▶ This algorithm has been instantiated in Soar/SVS.

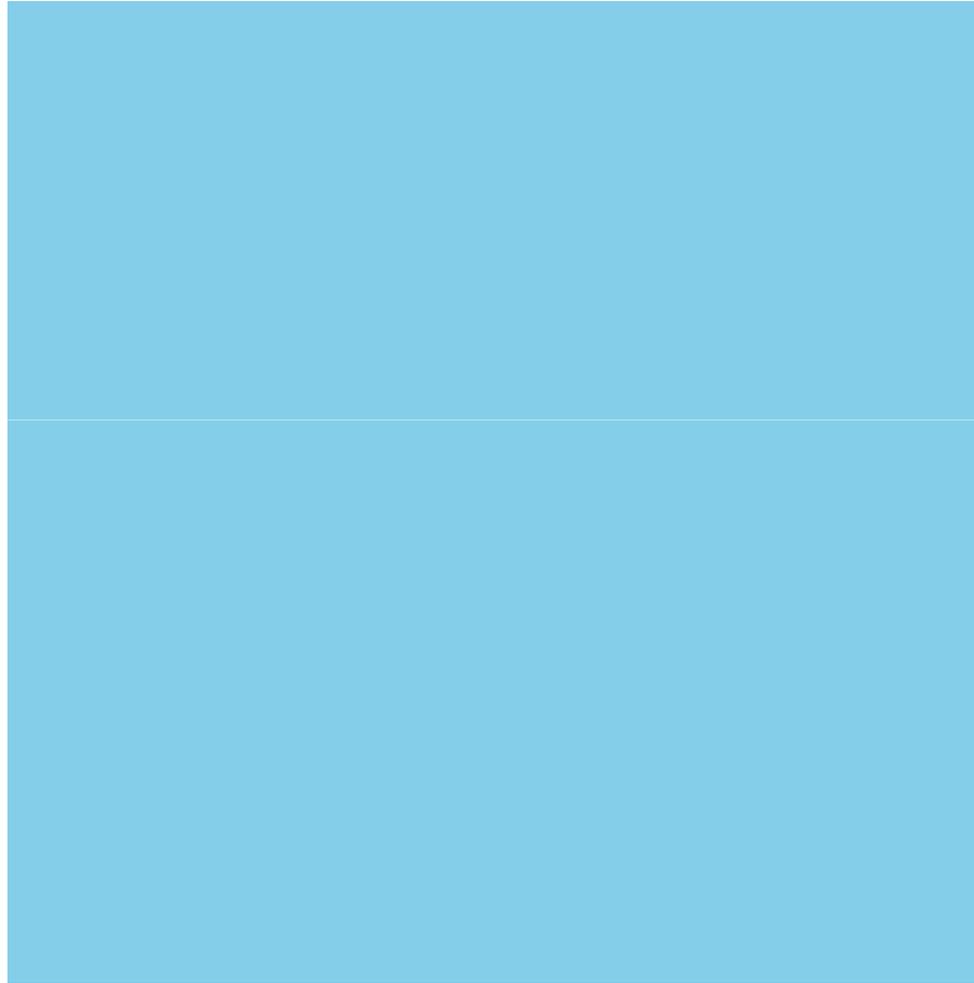
RRT Motion Planning

- ▶ Controller steers car toward a goal, biasing steering away from obstacles
- ▶ Soar keeps track of tree of possible configurations, and chooses which to expand next



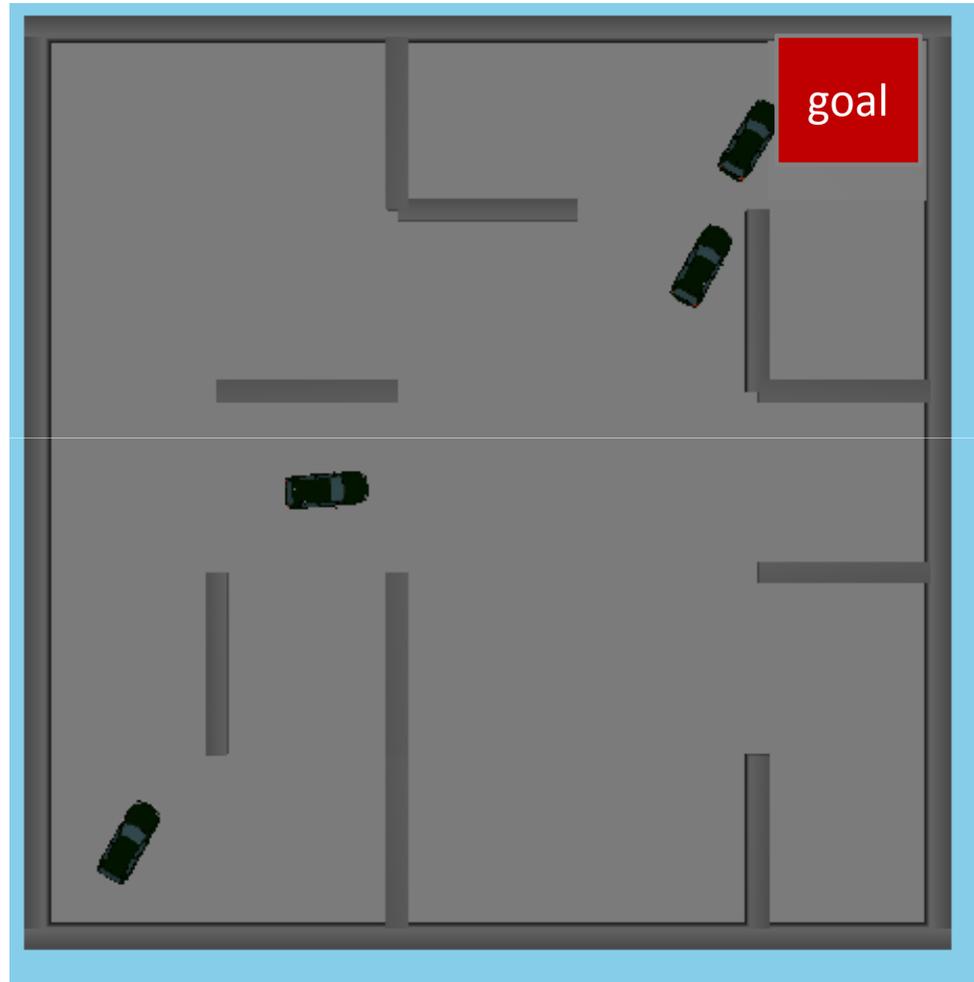
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High-Level Perceptions Used

- ▶ Pegged Blocks World:
 - ▶ **intersecting(X,Y); above(X,Y); distance(X,Y)**
 - ▶ Symbolically composed to `on(X,Y)` and `collision(X,Y)` predicates
- ▶ RRT Motion Planning:
 - ▶ **intersecting(X,Y); in-front-of(X,Y); distance(X,Y)**
 - ▶ Used to determine if car has hit an obstacle or reached the goal, and choose which node to expand
- ▶ Low-level perception and object recognition are (hypothetically) the same in both cases.

Imagery Actions Used

- ▶ Pegged Blocks World:
 - ▶ predicate projection: imagine a copy of a block on top of an existing block, centered relative to it
- ▶ RRT Motion Planning:
 - ▶ predicate projection: imagine a random point within the floor
 - ▶ motion simulation: imagine a copy of the car in the future, given its motion model

Conclusion

- ▶ Nuggets:
 - ▶ Imagery allows spatial problems to be accurately solved even when no good abstraction is available
 - ▶ Imagery reduces complicated perceptual operations to simple operations performed over time
 - ▶ This results in a small set of perceptual primitives
 - ▶ This allows the same perceptual system to be used in many problem domains
- ▶ Coal:
 - ▶ Low-level perception is still hard
 - ▶ Generality is a hard claim to evaluate
 - ▶ Still no software release

References

- ▶ SVS
 - ▶ S. Wintermute, *An Overview of Spatial Processing in Soar/SVS*, Technical Report, University of Michigan Center for Cognitive Architecture, 2009.
- ▶ Pegged blocks world agents
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 - ▶ S. Wintermute and J.E. Laird, “Imagery as Compensation for an Imperfect Abstract Problem Representation,” *Proceedings of the 31st Annual Conference of the Cognitive Science Society*, 2009.
- ▶ RRT algorithm
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- ▶ RRT agent
 - ▶ S. Wintermute, “Integrating Reasoning and Action through Simulation,” *Proceedings of the Second Conference on Artificial General Intelligence*, 2009.