

Representing and Reasoning with Intentional Actions on a Robot

PlanRob - Architectures and Frameworks
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Talk Outline

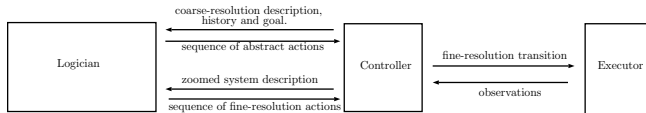
- Motivation and architecture overview.
- Knowledge representation and reasoning.
- Adapted Theory of Intentions.
- Experiments and conclusions.

Architecture Overview

Design architecture for robots that:

- Use **qualitative** and **quantitative** descriptions of incomplete knowledge to improve quality and efficiency of decision making.
“books are usually in the library”
“I am 90% certain the book is on the table”
- Work in complex dynamic domain where unexpected things happen.
- Receive more sensor data than it can handle at reasoning level.

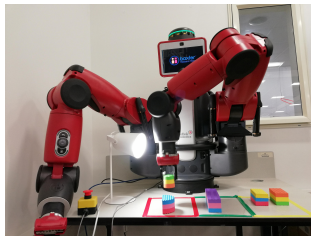
Architecture Overview



Architecture Overview

- Combine strengths of declarative programming and probabilistic models.
- **Formal relationship** between transition diagrams. **Tight coupling**, not a “switch”.
- Support **non-monotonic logic** and **probabilistic** representations of **incomplete domain knowledge** and **uncertainty**.
- **Interactive** and cumulative **relational learning**; classical and reinforcement, procedural and knowledge-based.
- Implemented on **robots** interacting with dynamic environments, and in **simulation**.

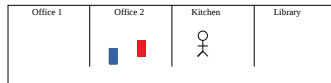
Illustrative Domain: Robot Moving Objects



- Robot **assistant** moving objects to specific places.
- **Sorts**: *robot, place, object, entity, book* etc; hierarchical arrangement.
- Instances, e.g., **place** = $\{office, kitchen, library\}$ or $\{red_area, green_area, yellow_area\}$ for tabletop manipulation.

Illustrative Domain: Robot Moving Objects

Let us assume that the robot is in the *kitchen* and its goal is to put the two books in the *library*.



In traditional planning the robot would:

- Create a plan as a sequence of actions, taking into consideration current beliefs and goal.
- Execute the actions of the sequence until one action does not have the expected output.
- Replan if necessary.

Illustrative Domain: Robot Moving Objects



In a dynamic domain we can have different scenarios:

- While the robot is moving the first book, the second book is moved to the library (**unexpected success**).
- While the robot is moving the first book, the second book is moved to office 1 (**not expected to achieve goal - case 1**).
- While the robot is moving the first book, the second book is moved to the kitchen (**not expected to achieve goal - case 2**).
- The robot moves the first book in the library, and when he is putting down the second book, the first book is being moved to another room (**failure to achieve the goal**).

Action Language AL_d

- Formal models of parts of natural language used for describing transition diagrams.
- Hierarchy of basic **sorts**, **statics**, **fluents** (basic and defined) and **actions**.
- Five types of **statements**:
 - Deterministic causal law.
 - Non-deterministic causal law.
 - State constraint.
 - Definition.
 - Executability condition.

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Logician's System Description

- Collection of statements of AL_d forms **system description** \mathcal{D}_H , includes **sorted signature** Σ_H and axioms.
- **Statics**: $next_to(kitchen, library)$.
- **Fluents**:
 $loc : robot \rightarrow place$
 $in_hand : robot \times book \rightarrow boolean$
- **Actions**:
 $move(robot, place)$
 $pickup(robot, book)$

Logician's System Description

- **Causal laws:**

$move(rob_1, Pl)$ **causes** $loc(rob_1) = Pl$

$pickup(rob_1, B)$ **causes** $in_hand(rob_1, B)$

- **State constraints:**

$loc(B) = Pl$ **if** $loc(rob_1) = Pl$, $in_hand(rob_1, B)$

$loc(Th) \neq Pl_1$ **if** $loc(Th) = Pl_2$, $Pl_1 \neq Pl_2$

- **Executability conditions:**

impossible $pickup(rob_1, O)$ **if** $loc(rob_1) \neq loc(O)$

impossible $pickup(rob_1, O1)$ **if** $in_hand(rob_1, O2)$

Logician's Reasoning

Refine + Zoom

- Examine system description (\mathcal{D}_H) at finer resolution (\mathcal{D}_L).
- Inherit knowledge; add **knowledge** fluents and actions.
- Automatically **zoom** to $\mathcal{D}_{LR}(T)$ relevant to $T = \langle \sigma_1, a^H, \sigma_2 \rangle$.
- **Formal relationships** between descriptions.

POMDP Construction

- $\mathcal{D}_{LR}(T)$ and statistics to construct **Partially Observable Markov Decision Process** (POMDP) tuple $\langle S^L, A^L, Z^L, T^L, O^L, R^L \rangle$.
- Add observed outcomes to \mathcal{H} to be used by logician.

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Adapted Theory of Intentions

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Experimental Results: Speed and accuracy

Advantages of Methodology

Conclusions + Future Work

- **Conclusions:**
- **Future Work:**

That's all folks!