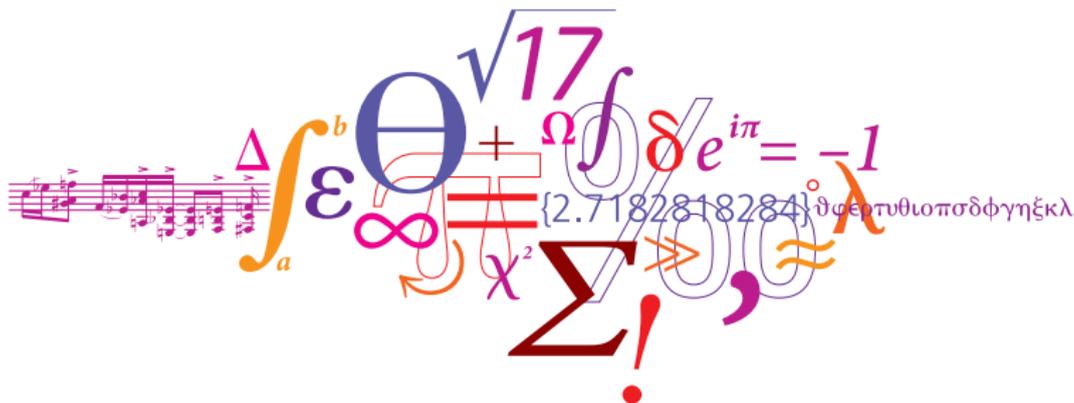
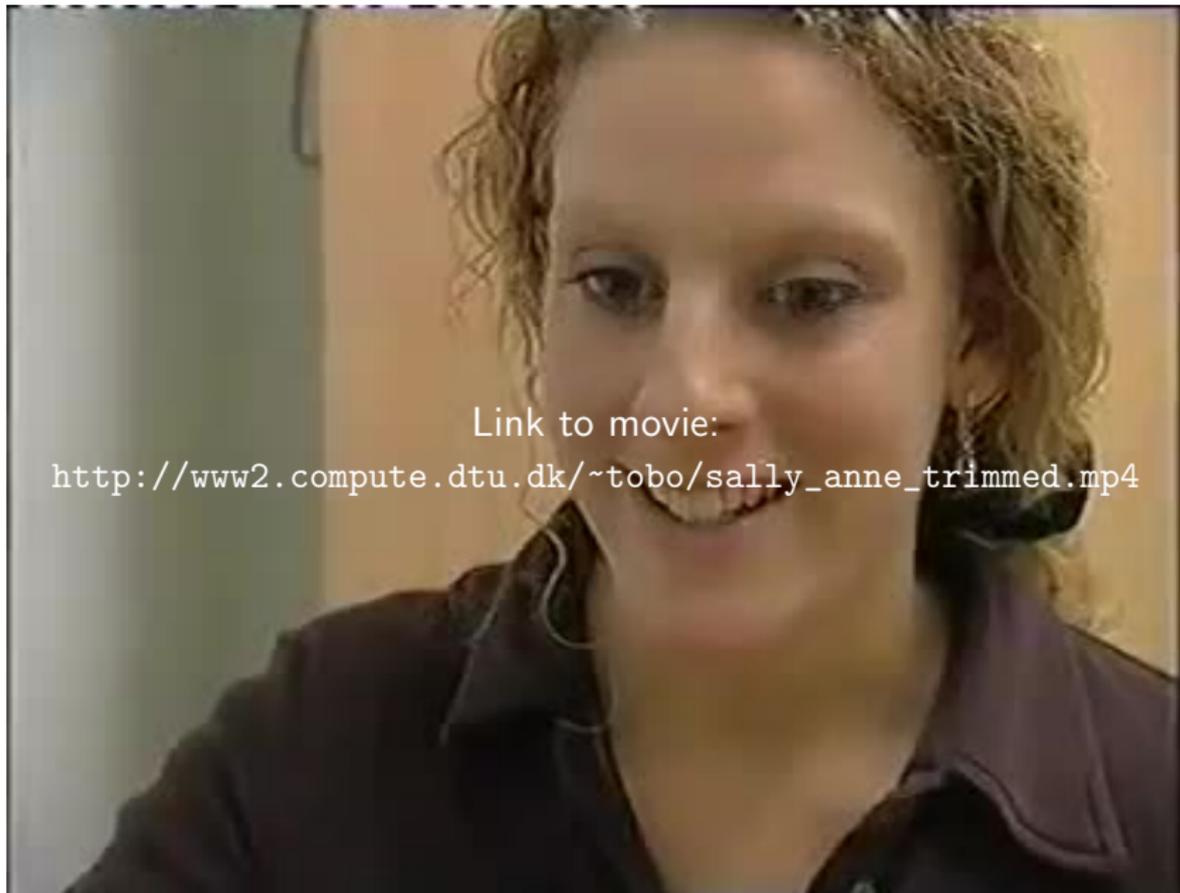


Theory of Mind and Epistemic Planning for Human-Robot Collaboration

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Testing Theory of Mind: the Sally-Anne test



Link to movie:

http://www2.compute.dtu.dk/~tobo/sally_anne_trimmed.mp4



Link to movie:

http://www2.compute.dtu.dk/~tobo/komdigital_pepper_video.mov

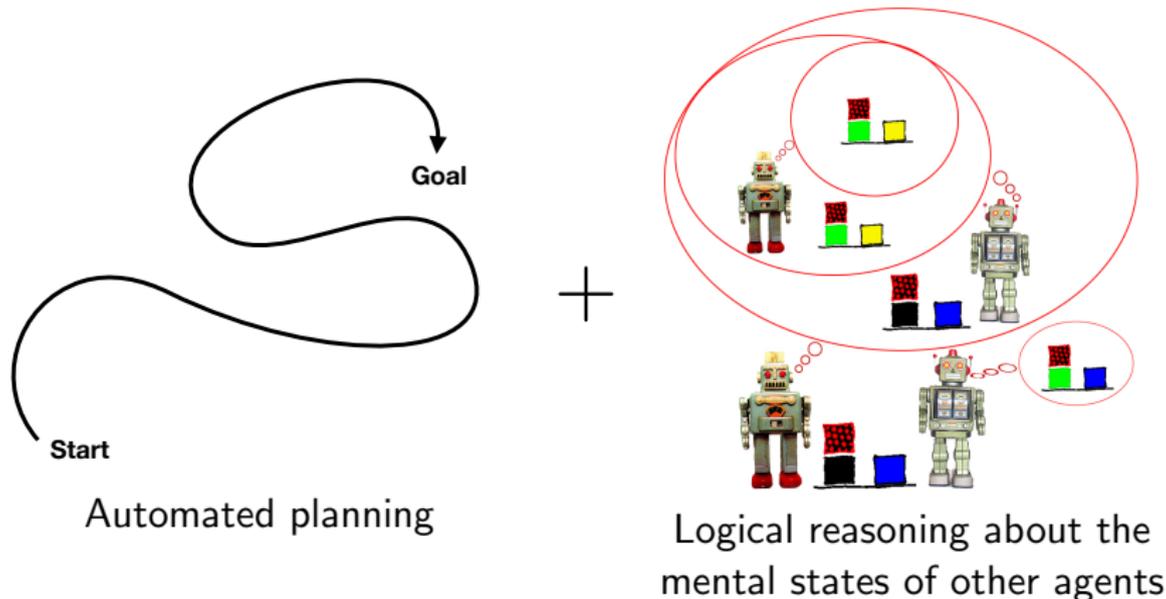
“R2DTU A Pepper robot with social intelligence”

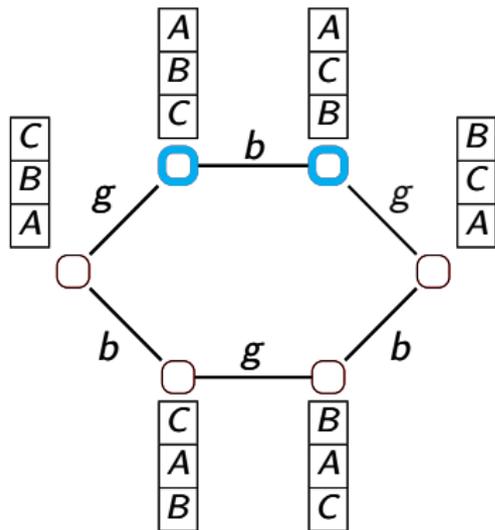
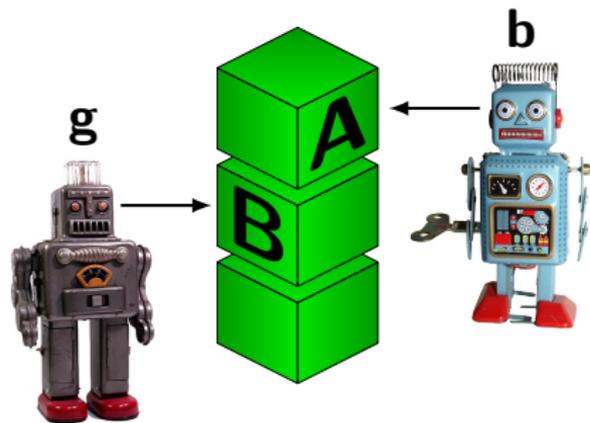
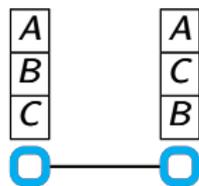
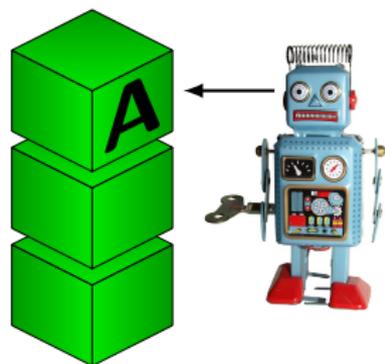
(KomDigital: R2DTU – A Pepper robot, 25 November 2020)

Epistemic planning =
automated *planning* + Theory of Mind reasoning

Aim: To compute plans that can take the mental states of other agents into account.

Essentially: (Decentralised) **multi-agent planning** in environments with (potentially higher-order) **information asymmetry**.





Epistemic states: Multi-pointed epistemic models of multi-agent S5. Nodes are worlds. **Designated worlds:** ○ (those considered possible by planning agent).

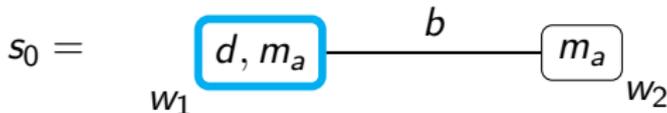
The coordinated attack problem in dynamic epistemic logic (DEL)

Two generals (agents), a and b . They want to coordinate an attack, and only win if they attack simultaneously.

d : “general a will attack at dawn”.

m_i : the messenger is at general i (for $i = a, b$).

Initial **epistemic state**:

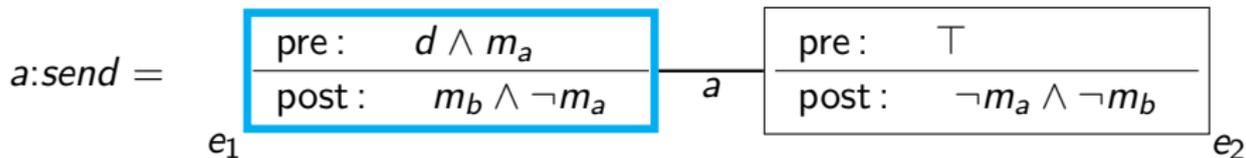


Nodes are **worlds**, edges are **indistinguishability edges** (reflexive loops not shown).

The coordinated attack problem in dynamic epistemic logic (DEL)

Recall: d means “ a attacks at dawn”; m_i means messenger is at general i .

Available **epistemic actions** (aka **action models** aka **event models**):

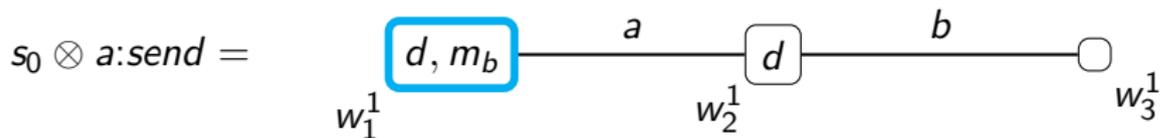
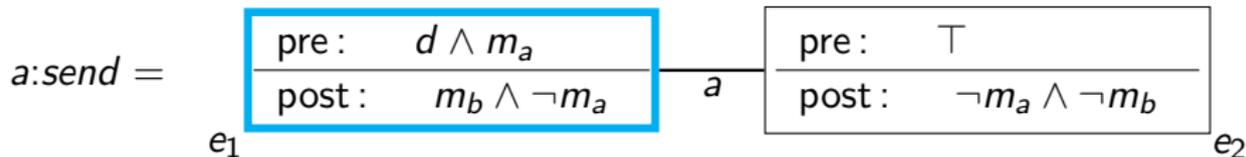
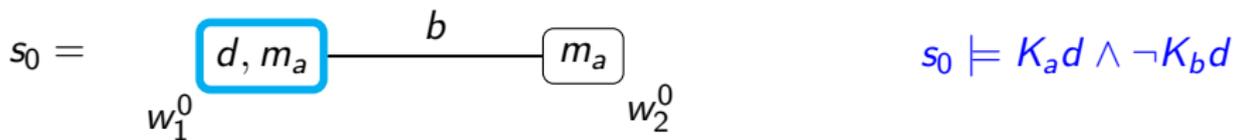


And symmetrically an epistemic action $b:send$. We read $i:\alpha$ as “agent i does α ”.

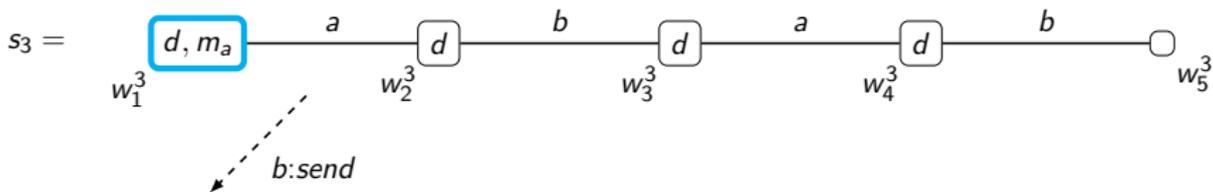
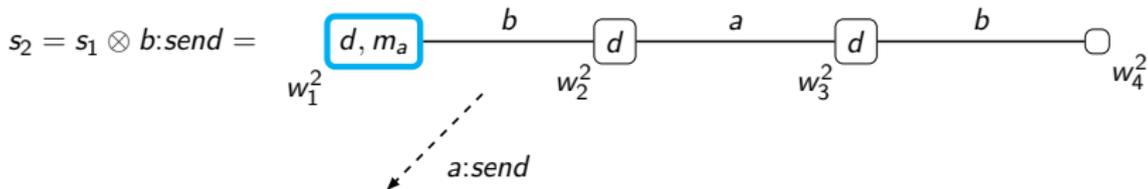
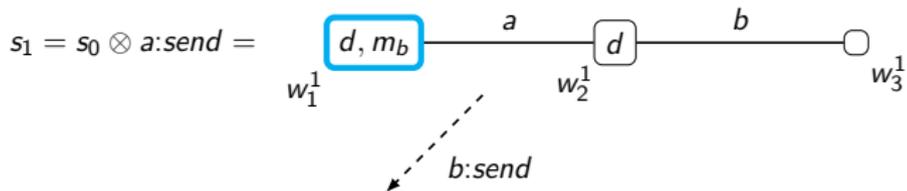
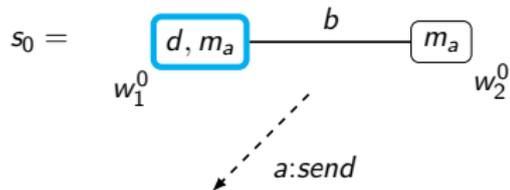
Nodes are **events**, and each event has a **precondition** and a **postcondition** (effect). The precondition is an epistemic formula and the postcondition is a conjunction of literals.

[Baltag et al., 1998, van Ditmarsch and Kooi, 2008]

The product update in dynamic epistemic logic



$s_0 \otimes a:send \models K_a d \wedge K_b d \wedge \neg K_a K_b d$



Epistemic planning tasks

Definition. An **epistemic planning task** (or simply a **planning task**) $T = (s_0, A, \gamma)$ consists of an epistemic state s_0 called the **initial state**; a finite set of epistemic actions A ; and a **goal formula** γ of the epistemic language.

Definition. A (sequential) **solution** to a planning task $T = (s_0, A, \gamma)$ is a sequence of actions $\alpha_1, \alpha_2, \dots, \alpha_n$ from A such that for all $1 \leq i \leq n$, α_i is applicable in $s_0 \otimes \alpha_1 \otimes \dots \otimes \alpha_{i-1}$ and

$$s_0 \otimes \alpha_1 \otimes \alpha_2 \otimes \dots \otimes \alpha_n \models \gamma.$$

Example. Let s_0 be the initial state of the coordinated attack problem. Let $A = \{a:send, b:send\}$. Then the following are planning tasks:

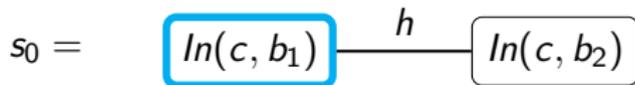
1. $T = (s_0, A, Cd)$, where C denotes common knowledge. It has no solution.
2. $T = (s_0, A, E^n d)$, where E denotes “everybody knows” and $n \geq 1$. It has a solution of length n .

[Bolander et al., 2020]

Epistemic planning example: Get the cube

- **Objects:** $\mathcal{O} = \{b_1, b_2, c\}$, two boxes b_1 and b_2 , and a cube c .
- **Agents:** $\mathcal{A} = \{h, a\}$, a human h and a robot r . The robot is the planning agent.
- **Atomic propositions:** $In(x, y)$ means x is in y , where $x, y \in \mathcal{O} \cup \mathcal{A}$ (when $y \in \mathcal{A}$, it means y is holding x).

Initial epistemic state:



The goal is for the human to hold the cube, $In(c, h)$.

Actions specialised for the case of $\mathcal{O} = \{b_1, b_2, c\}$.

Agent i (semi-privately) **peeks** into box x :

$$i:\text{peek}(x) = \boxed{\text{pre: } \text{In}(c, x)} \xrightarrow{\mathcal{A} - \{i\}} \boxed{\text{pre: } \neg\text{In}(c, x)}$$

Agent i (publicly) **picks up** object x from y :

$$i:\text{pickup}(x, y) = \boxed{\frac{\text{pre: } \text{In}(x, y)}{\text{post: } \text{In}(x, i) \wedge \neg\text{In}(x, y)}}$$

Agent i (publicly) **puts** object x in y :

$$i:\text{putdown}(x, y) = \boxed{\frac{\text{pre: } \text{In}(x, i)}{\text{post: } \text{In}(x, y) \wedge \neg\text{In}(x, i)}}$$

Agent i (publicly) **announces** that formula φ is true:

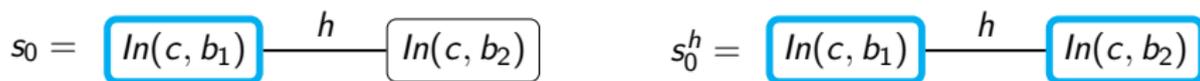
$$i:\text{ann}(\varphi) = \boxed{\text{pre: } \varphi}$$

Applicability, perspective shifts, implicit coordination

Seemingly simpler solution: $h:pickup(c, b_1)$. But intuitively, this shouldn't work, since the human doesn't know the cube is in box 1...

Applicability: An action α is **applicable** in a state s if for each designated world w of s there is a designated event e of α with $w \models pre(e)$.

Perspective shift: The **perspective shift** of state s to agent i , denoted s^i , is achieved by closing under the indistinguishability relation of i . We call s^i the **perspective** of agent i on state s .

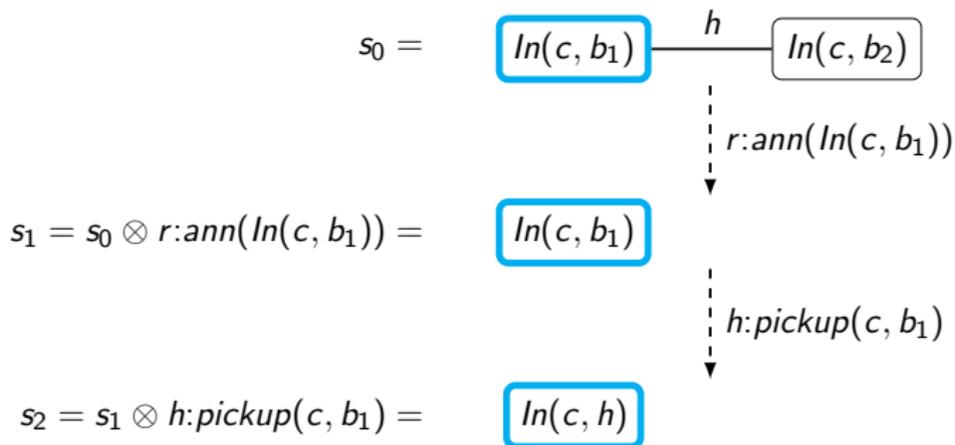


Example. $h:pickup(c, b_1)$ is not applicable in s_0 from h 's perspective.

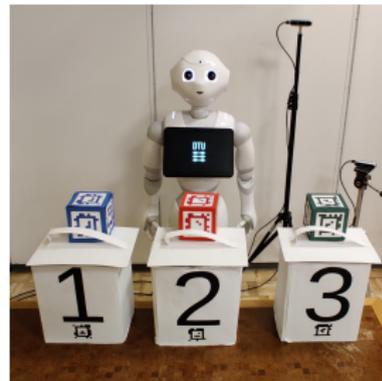
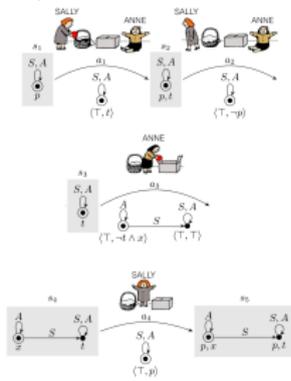
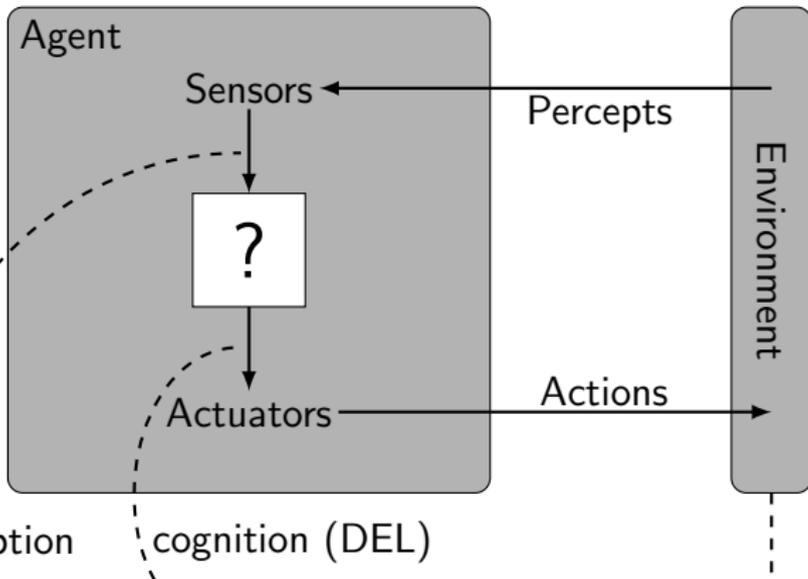
Implicitly coordinated solution to planning task: Each action has to be applicable from the perspective of the acting agent; and the product update $s \otimes i:\alpha$ is replaced by $s^i \otimes i:\alpha$.

Get the cube: Implicit coordination

Joint solution to T , by robot R , implicitly coordinated:



If purely epistemic actions (announcements) have a lower cost than ontic actions (moving things around), the solution above is the only optimal one.



Perception layer: detectors, world model and events

Detectors: Detect a specific kind of feature such as faces (dlib CNN face recognition), markers (AprilTag fiducial markers), and body poses (OpenPose).

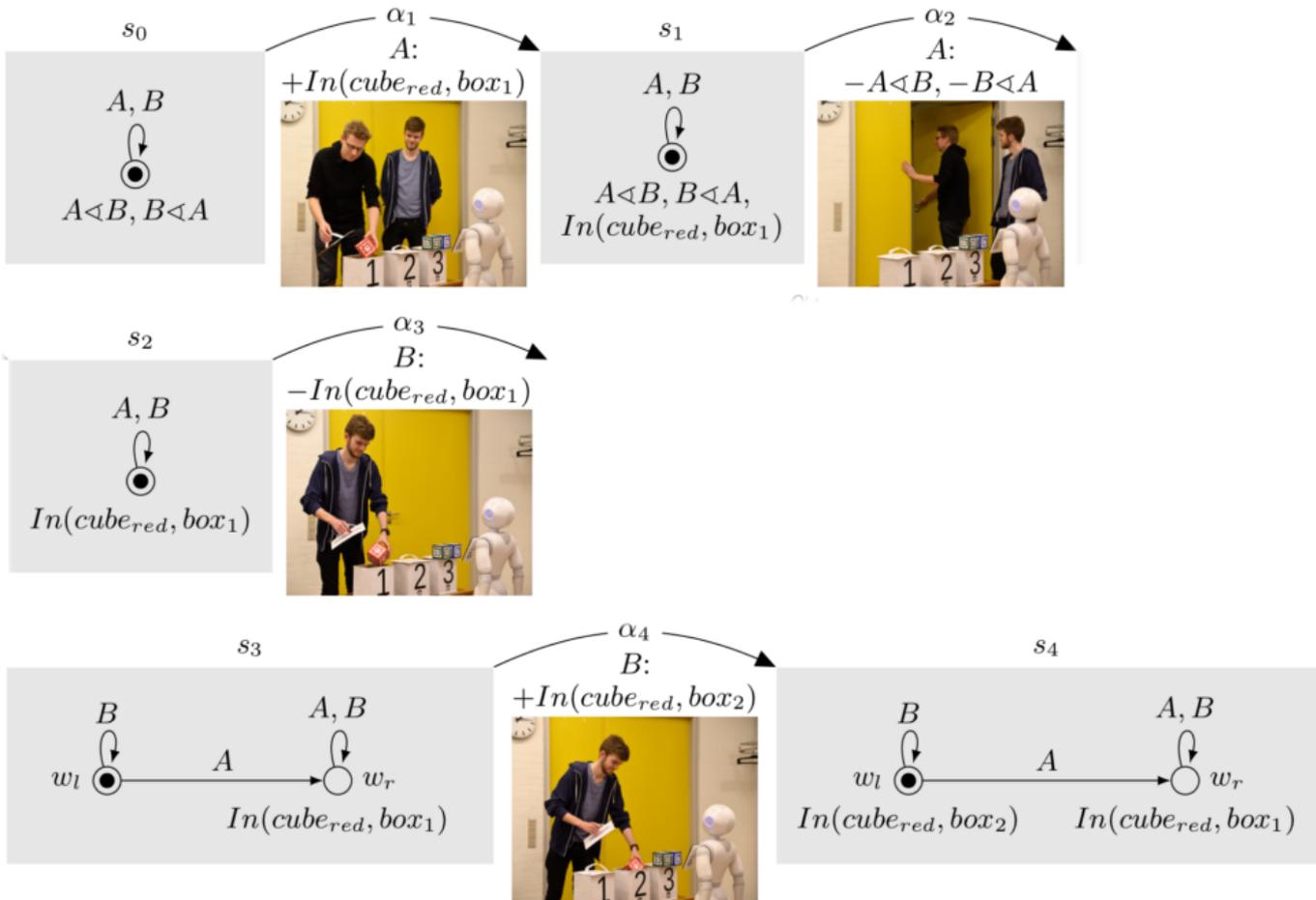
Spatial world model: Keeps track of the spatial position of physical entities using the detectors. Physical entities are split into *objects* \mathcal{O} and *agents* \mathcal{A} .

Events: Changes in the spatial world model triggers *events*:

- $\text{Appear}(c)/\text{Disappear}(c)$: World model starts/stops tracking entity c .
- $\text{pickup}(i, c)$: Agent i picks up object c . Triggered by hand of i entering bounding box of c .
- $\text{put}(i, c, b)$: Agent i puts object c in container b .

From perception layer to cognition layer: Every event is translated into its corresponding epistemic action and applied to the current epistemic state via the product update.

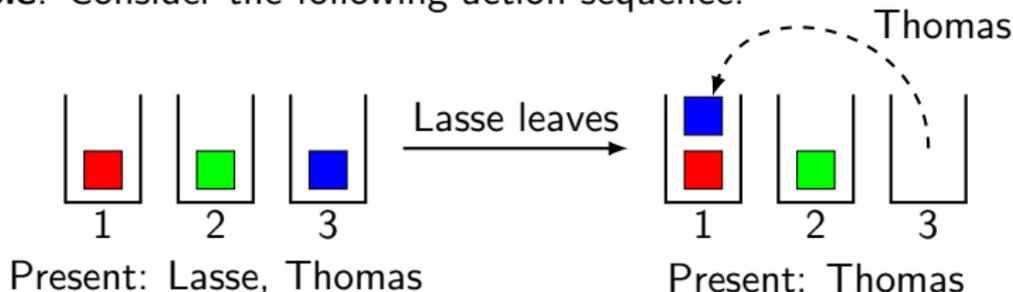
E.g. $\text{put}(i, c, b) \curvearrowright i:\text{putdown}(c, b)$.



Helpful announcements

- We add announcements, so the robot can be helpful by announcing facts.
- The robot does epistemic planning with implicit coordination: multi-agent planning with perspective shifts [Nebel et al., 2019, Bolander et al., 2018, Engesser et al., 2017].

Example. Consider the following action sequence:



If I say “I want two cubes in the same box”, nothing happens. Lasse arrives and says the same. Now the robot replies: “It is already true”.

Afterwards Lasse says: “I want three cubes in the same box”. The robot replies: “Box 3 is empty”.

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