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A4M33MAS - Multiagent Systems

Intelligent agents architectures

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In parts based on Michal Pechoucek: Multi-Agent Systems, Lecture course at State University of New York, University at Binghamton

Towards Architectures for IA

01

- Reactive Architectures
- Deliberative Architectures

Subsumption Architecture

best known reactive agents architecture developed by R. Brooks:

- Two key ideas:
 - **situatedness** and **embodiment** of intelligence: real intelligence is situated in the world it is not disembodied such as expert system or theorem prover
 - **intelligence** and **emergence**: intelligence arises from the agent x environment interaction, intelligence is not an isolated property
- **Task accomplishing behavior**: a function that maps the percept into action as there is no representation/reasoning the task accomplishing behaviors are implemented as rules:

`situation → action`

- the rules are fired simultaneously – the rules are structures into **subsumption hierarchy**: layers of different levels of abstraction of the behavior (where each layer inhibit the higher level layer)

Subsumption Architecture

- behavior:

$$\text{Beh} = \{\text{cond} \rightarrow \text{act}, \text{ where } \text{cond} \subseteq \mathcal{P} \text{ and } \text{act} \subseteq \mathcal{Ac}\}$$

- inhibition relation:

- for rules in Beh, there is a binary relation – a strict total ordering on Beh (i.e. transitive, irreflexive and antisymmetric)

```

fired → {cond → act, where cond → act ∈ P and act ⊆ Ac}
for each cond → act ∈ fired do
  if ∄ cond' → act' ∈ fired such that cond' → act' < cond → act
    then return act
  end-if
end-for
return null

```

Subsumption Architecture

- Example of rock sample collecting robots (Steels), inspired by ants:
 - noncooperative agents

```
(cond (detect-an-obstacle #'change-direction)      r1
      ((and carrying-samples at-base) #'drop-samples)  r2
      ((and carrying-samples (not at-base)) #'travel-up) r3
      ((detect-a-sample #'pick-up-sample)           r4
      (t #move-randomly)                            r5
      )
```

$r1 \prec r2 \prec r3 \prec r4 \prec r5$

Subsumption Architecture

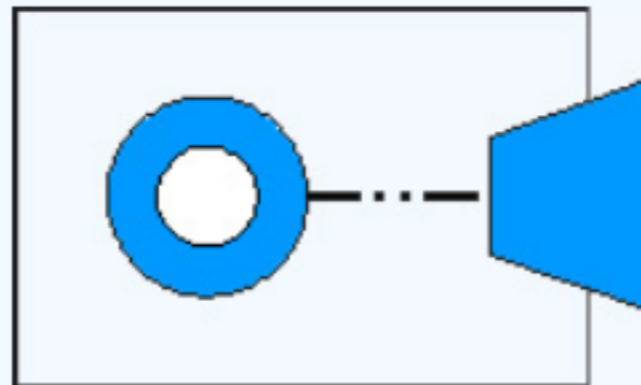
- if samples in clusters, collaborative property is desirable
 - direct/indirect communication
- **stigmergy**: ant-inspired collaborative decision making

```
((and carrying-samples (not at-base))  
    (and #drop-2-crumbs #'travel-up))          r6  
((detect-crumbs (and #'pick-up-1-crumb #'travel-down))  r7  
  
    r1 < r2 < r6 < r4 < r7 < r5
```

Subsumption Architecture

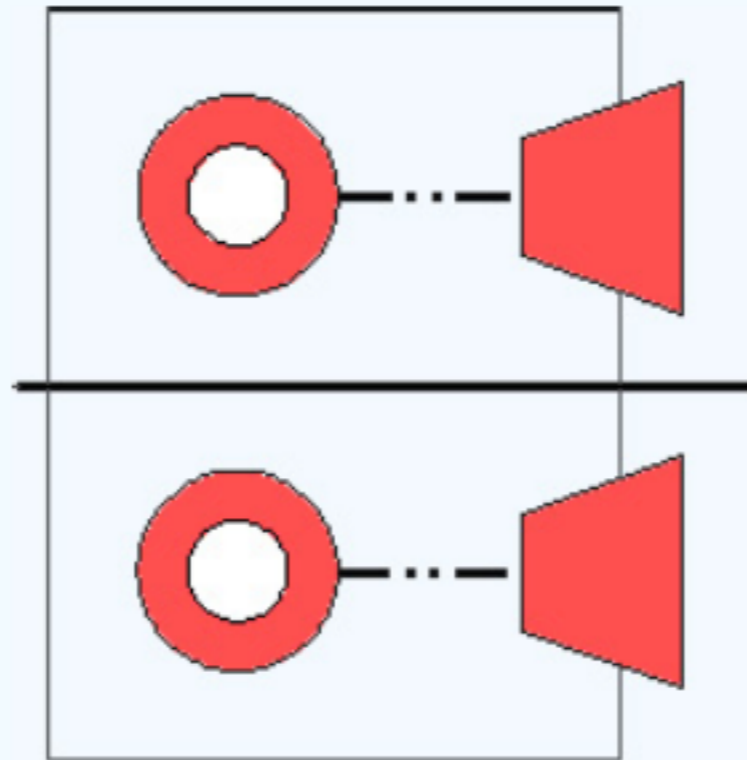
- S-agents

```
(cond  
  (light-in-front-increasing #move-forward-faster)  
)
```



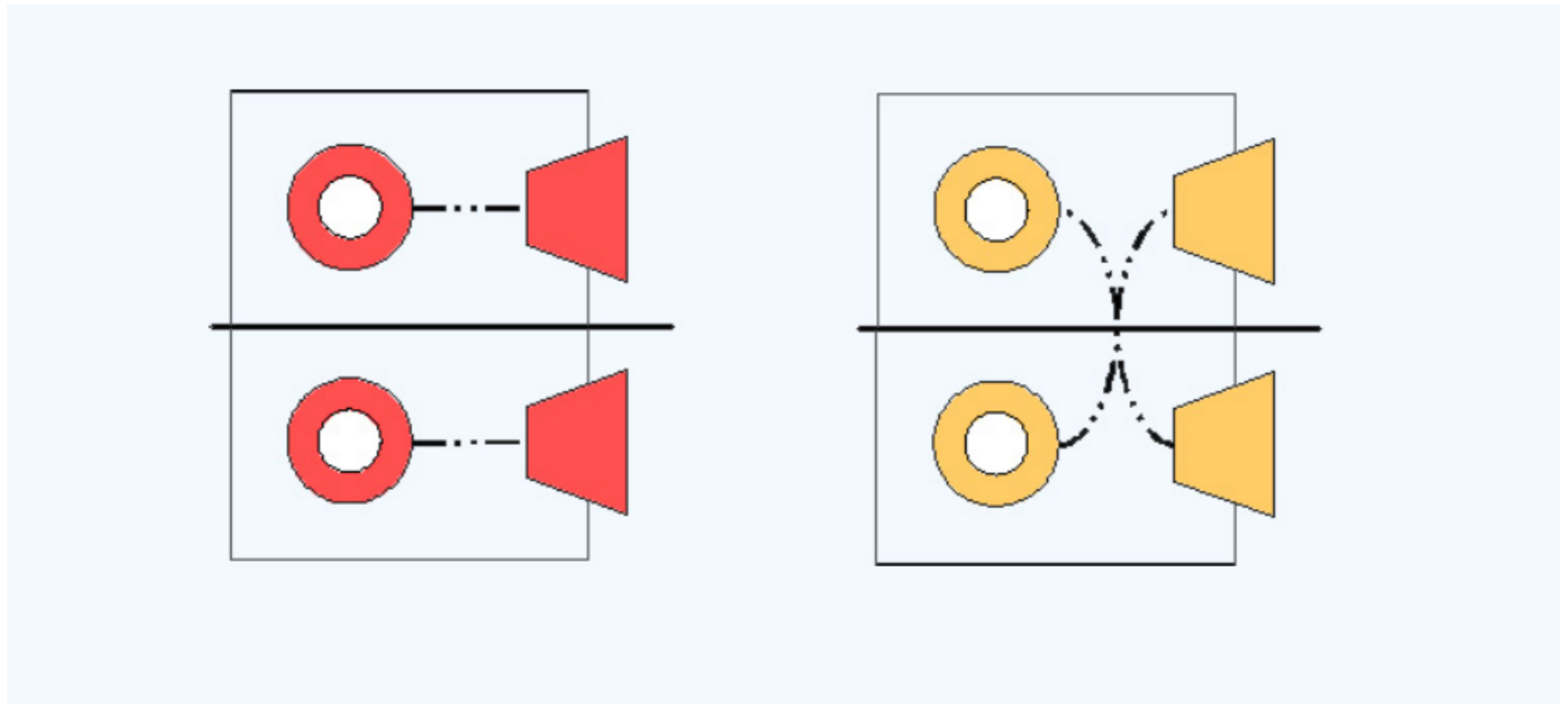
Subsumption Architecture

- S-agents: composition



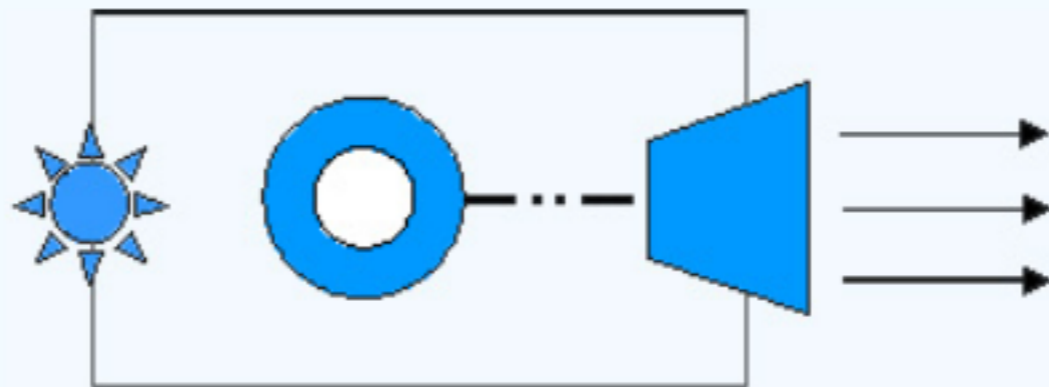
Subsumption Architecture

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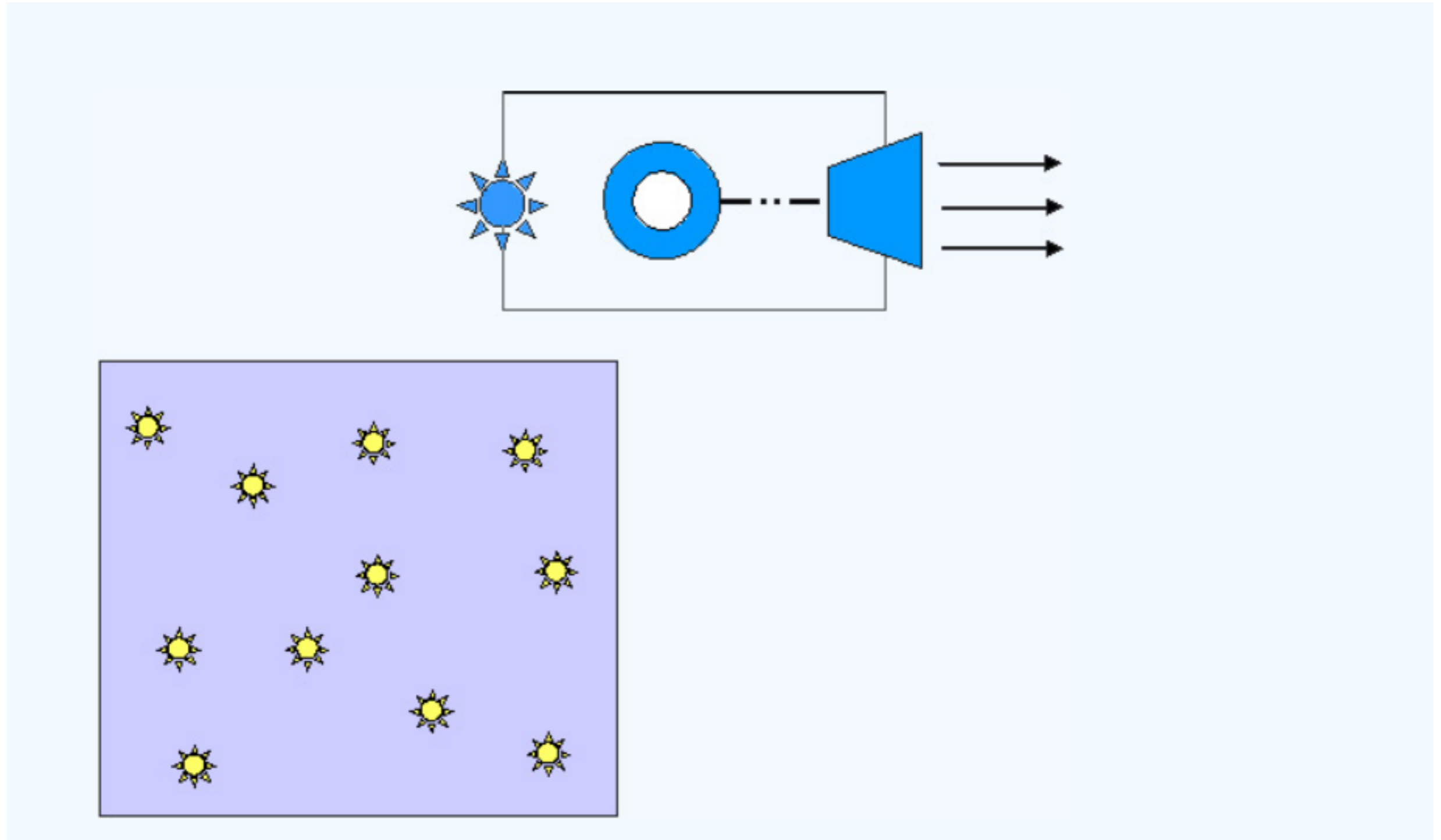
Subsumption Architecture

- S-agents: composition



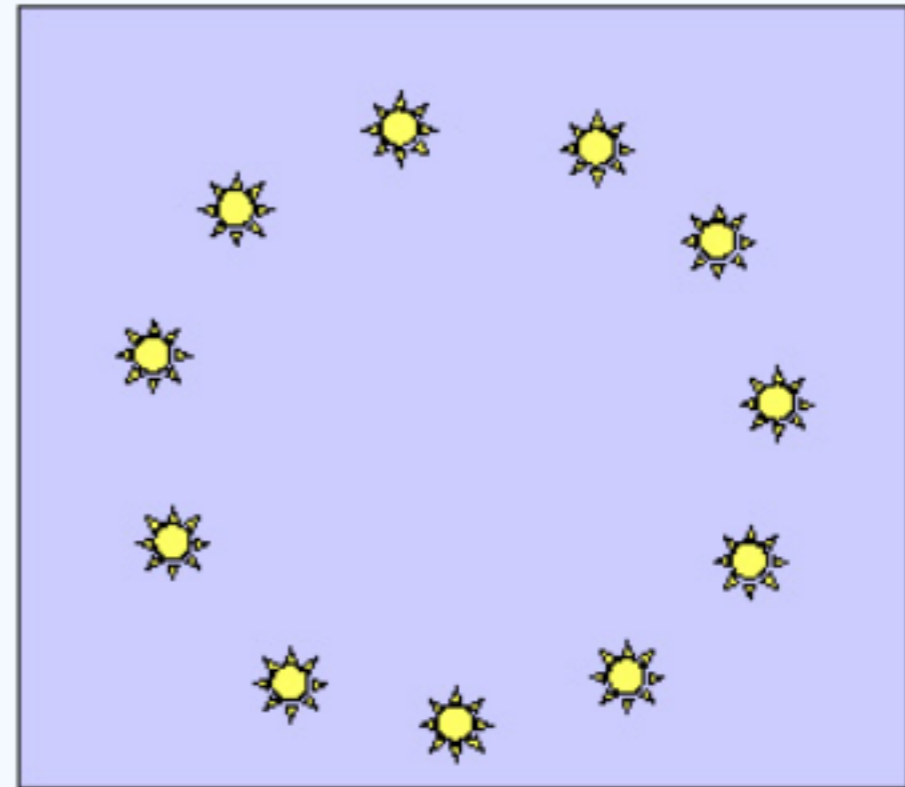
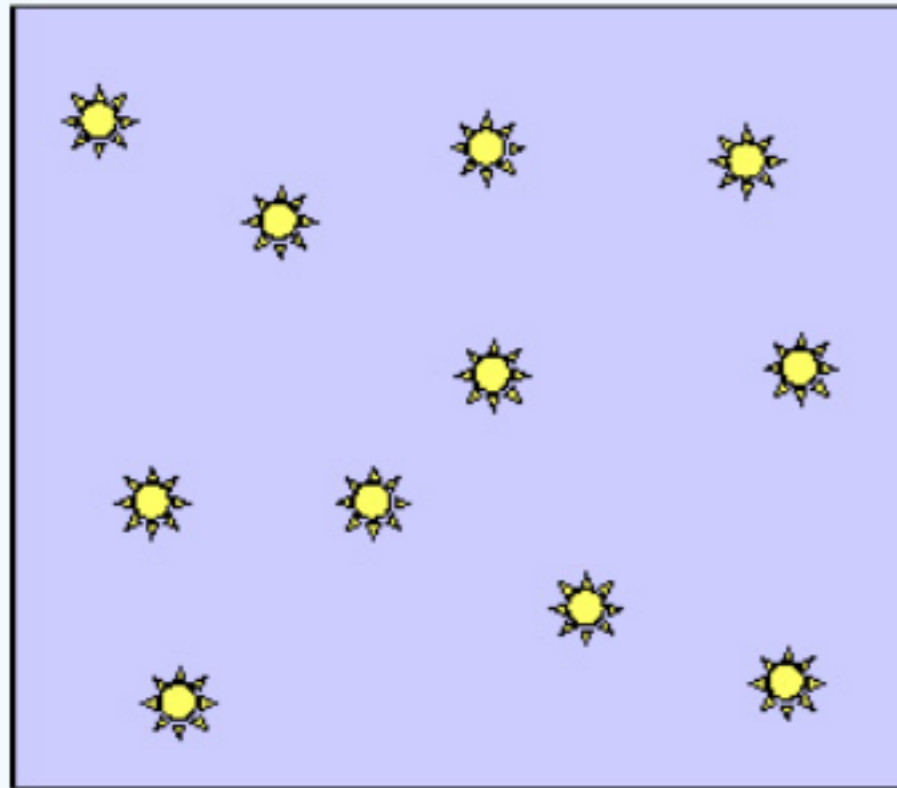
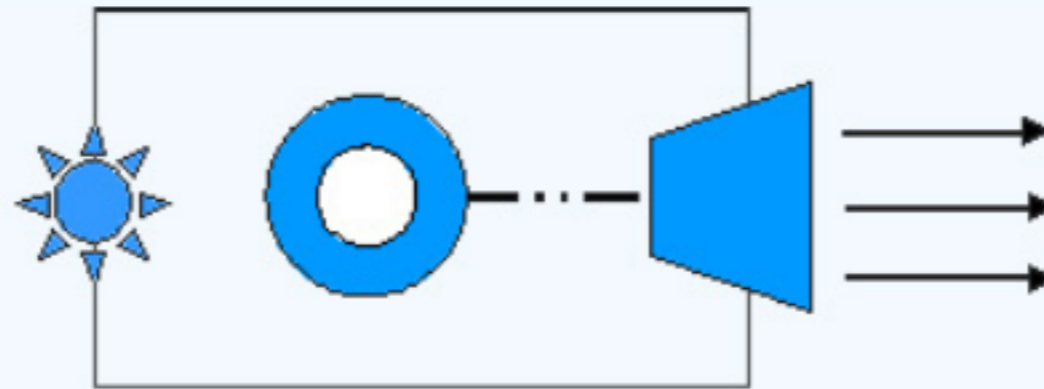
Subsumption Architecture

- S-agents: composition



Subsumption Architecture

- S-agents: composition



Deliberative architectures

01

1. Planning architectures
2. Deductive architectures
3. Models of practical reasoning

Planning Agents

01

Planning Agents

- Basic architecture:

1. Plan(s_0, s_g, O) $\rightarrow \pi$

2. Execute (π) $\rightarrow s_g$

Planning Agents

- **Basic architecture:**
 1. Plan(s_0, s_g, O) $\rightarrow \pi$
 2. Execute (π) $\rightarrow s_g$
- **Reactive planning architecture:**
 1. $s_0 \rightarrow s_n$
 2. Plan(s_n, s_g, O) $\rightarrow \pi$
 3. Read(Execute(*Head*(π))) $\rightarrow s_n$
 4. if $s_n \neq s_g$ *GOTO* 2.

Planning Agents

- **Basic architecture:**
 1. Plan(s_0, s_g, O) $\rightarrow \pi$
 2. Execute (π) $\rightarrow s_g$
- **Reactive planning architecture:**
 1. $s_0 = s_n$
 2. Plan(s_n, s_g, O) $\rightarrow \pi$
 3. Read(Execute(*Head*(π))) $\rightarrow s_n$
 4. if $s_n \neq s_g$ *GOTO* 2.

* For Plan see to A4M36PAH or related

Deductive Agents

- Models of deductive reasoning:

- constantly running a **reasoning loop** that is try to prove that there is an action that shall, or at least is allowed to be carried out – given agent's knowledge Δ , and its goal \mathcal{G} we shall instantiate a variable α in a predicate $\text{do}(\alpha)$ when proving $\Delta, \mathcal{G} \vdash \exists \alpha \text{ do}(\alpha)$ – where $\text{shall-be-done}(\alpha) \vee \text{may-be-done}(\alpha) \Rightarrow \text{do}(\alpha)$
- or prove what is the **best reaction** to a new piece of information \mathcal{P} sensed from the environment $\Delta, \mathcal{G} \vdash \exists \alpha \text{ react}(\mathcal{P}, \alpha)$

Reasoning rules are here the rules of mathematical logic. The agent's knowledge Δ contain agents' model of the environment and its (if-then) decision making rules.

Deductive Agents

- Implementation of deductive agents:

:: Firstly,

we shall find the proper logical language (e.g. FOPL) for representing agents model of its environment and than design a solid system that will represent the environmental objects that the agent may model, relation between them etc.


:: Secondly,

you have to design and implement/integrate an inference engine that will try compute (and instantiate) the logical consequence of the model in the $\text{do}(\alpha)$ sense.

We will need a theorem proving system, e.g. based on resolution technique (implemented in Prolog, Otter, etc.)

Deductive Agents

- Example

[0,2]	[12]	[22]
[01]	[11] 	[21]
[00]	[10]	[20]

We use 3 domain predicates in this exercise:

- $\text{In}(x, y)$ agent is at x, y
- $\text{Dirt}(x, y)$ there is dirt at x, y
- $\text{Facing}(d)$ the agent is facing direction d

Possible actions:

- $A_c = \{\text{turn, forward, suck}\}$

Deductive Agents

- Example

$$\text{In}(x, y) \wedge \text{dirt}(x, y) \rightarrow \text{do}(\textit{suck})$$
$$\text{In}(x, 0) \wedge \text{Facing}(\textit{north}) \wedge \neg \text{dirt}(x, 0) \rightarrow \text{do}(\textit{forward})$$
$$\text{In}(x, 0) \wedge \neg \text{Facing}(\textit{north}) \wedge \neg \text{dirt}(x, 0) \rightarrow \text{do}(\textit{turn}) \wedge \text{do}(\textit{forward})$$
$$\text{In}(x, y) \wedge \neg \text{dirt}(x, y) \rightarrow \text{do}(\textit{forward})$$
$$\text{In}(x, 2) \wedge \text{Facing}(\textit{south}) \wedge \neg \text{dirt}(x, 2) \rightarrow \text{do}(\textit{forward})$$
$$\text{In}(x, 2) \wedge \neg \text{Facing}(\textit{south}) \wedge \neg \text{dirt}(x, 2) \rightarrow \text{do}(\textit{turn}) \wedge \text{do}(\textit{forward})$$
$$\text{In}(2, 2) \wedge \text{do}(\textit{finish})$$

Problems with Deductive Agents

01

:: **calculative rationality (CR) requirements**

- an agent comply with calculative rationality requirements provided:

$$\forall \Delta_{1,2} : \Delta_1 \vdash \text{do}(\alpha_1) \wedge \Delta_2 \vdash \text{do}(\alpha_2) \Rightarrow \text{time}(\Delta_1 \rightsquigarrow \Delta_2) > \text{time}(\Delta_1 \vdash \text{do}(\alpha_1))$$

:: **first order logic is not expressive enough**

- we need mechanisms for expressing functions, effects and dynamics of actions, higher level modalities such as time, obligation, knowledge and agents mutual mental positions
- all this can be done in different logics (e.g. temporal, deontic, dynamic) but automated theorem proving in such systems is very, very complex and CR requirements are likely to fail

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* *but we have modal logic*



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