

Pervasive Computing

Ingegneria e Scienze Informatiche - UNIBO

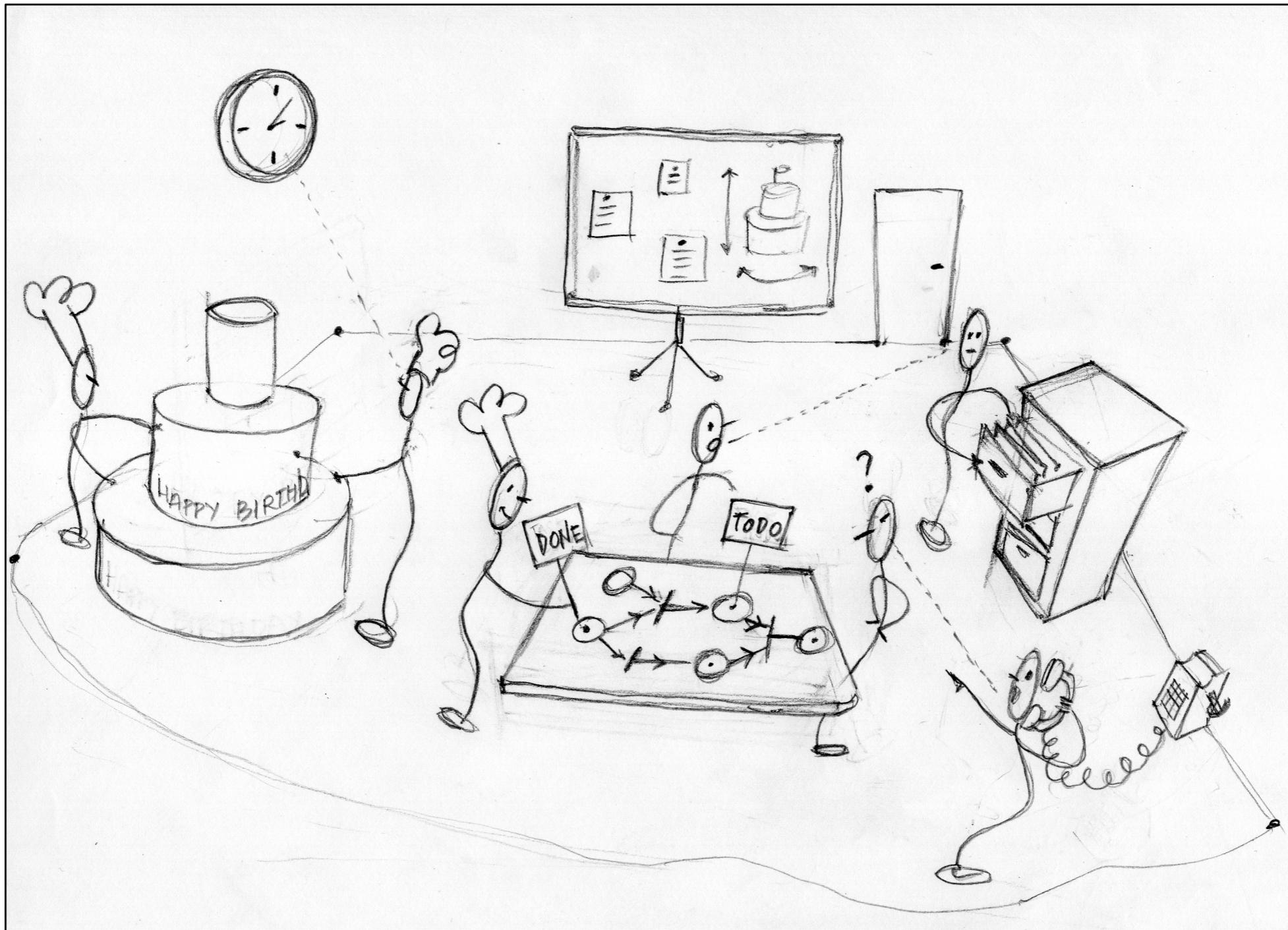
a.a 2017/2018

Lecturer: Prof. Alessandro Ricci

[PART THREE]
THEME/ISSUE:
AGENTS - BRIEF INTRO

DESIGNING AGENTS' WORLD

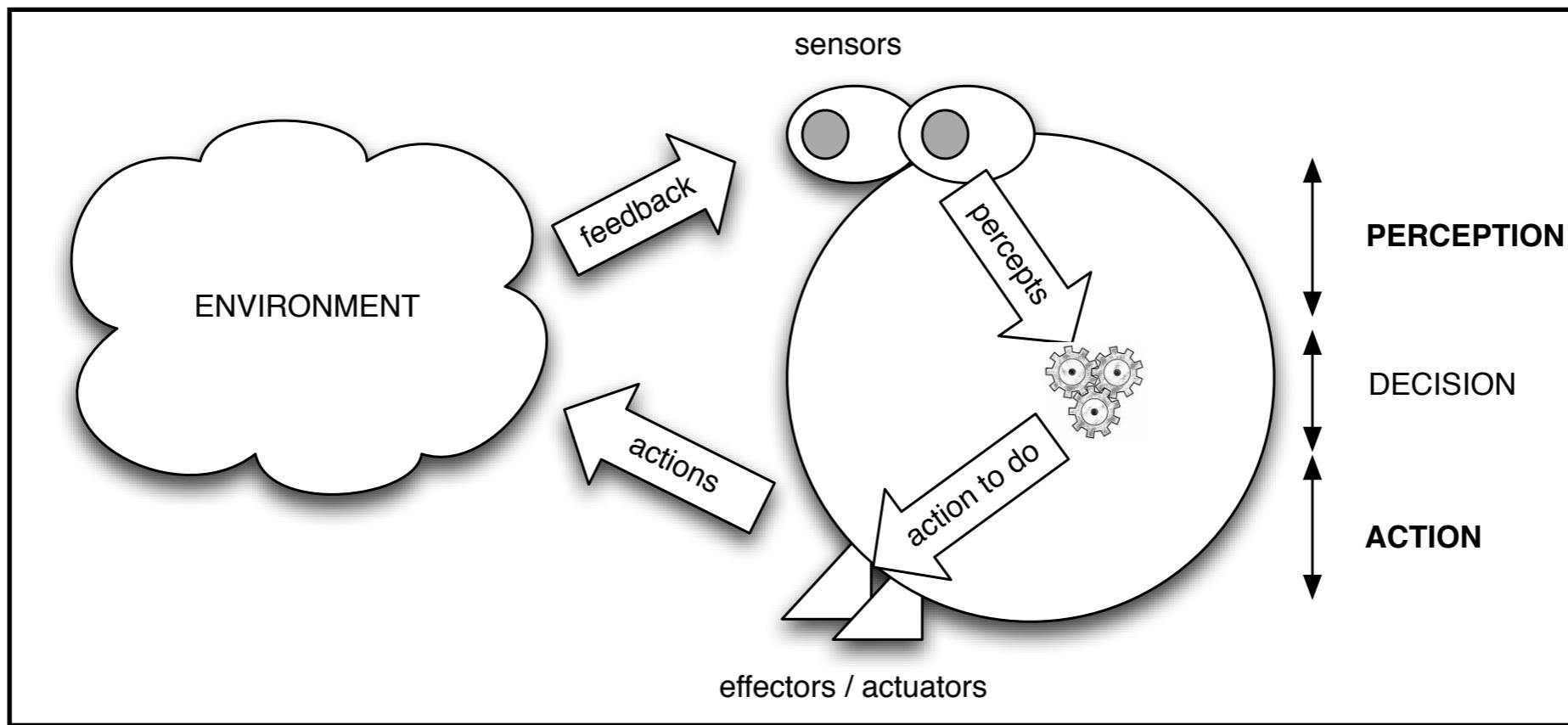
- A HUMAN METAPHOR -



THE NOTION OF AGENT

- Agent definition
 - “*An agent is a computer system that is situated in some environment and that is capable of autonomous action in this environment in order to meet its design objective*” [Wooldridge & Jennings, 1995]
 - “*An agent is anything that can be viewed as perceiving its environment through sensors and acting upon the environment through effectors.*” [Russell & Norvig, 1995]
- Including both physical and software (“computational”, “virtual”) environments

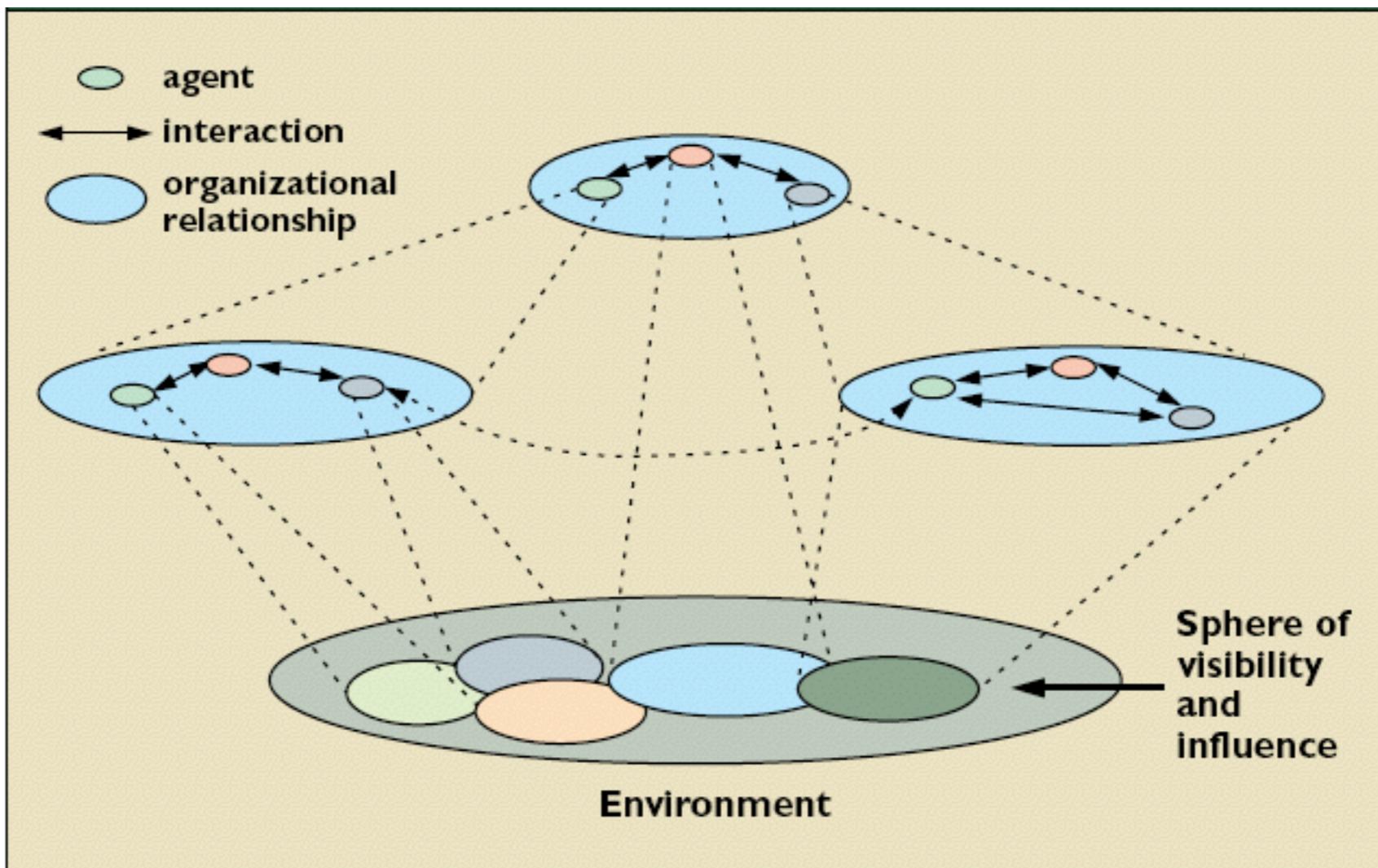
SINGLE-AGENT PERSPECTIVE



[Wooldridge, 2002]

- **Perception**
 - process inside agent inside of attaining awareness or understanding sensory information, creating **percepts**
 - perceived form of external stimuli or their absence
- **Actions**
 - the means to affect (change or inspect) the environment

MULTI-AGENT PERSPECTIVE



[Jennings, 2001]

- In evidence
 - spheres of visibility and influence
 - overlapping => interaction

ABSTRACT “ENVIRONMENT PROGRAM”

```
procedure RUN-ENVIRONMENT(state, UPDATE-FN, agents, termination)
  inputs: state, the initial state of the environment
          UPDATE-FN, function to modify the environment
          agents, a set of agents
          termination, a predicate to test when we are done

  repeat
    for each agent in agents do
      PERCEPT[agent] ← GET-PERCEPT(agent, state)
    end
    for each agent in agents do
      ACTION[agent] ← PROGRAM[agent](PERCEPT[agent])
    end
    state ← UPDATE-FN(actions, agents, state)
  until termination(state)
```

[Russell and Norvig, 1993]

A BASIC CLASSIFICATION

- *Accessible* versus *inaccessible*
 - indicates whether the agents have access to the complete state of the environment or not
- *Deterministic* versus *nondeterministic*
 - indicates whether a state change of the environment is uniquely determined by its current state and the actions selected by the agents or not
- *Static* versus *dynamic*
 - indicates whether the environment can change while an agent deliberates or not
- *Discrete* versus *continuous*
 - indicates whether the number of percepts and actions are limited or not

FURTHER CLASSIFICATION

[Ferber, 1999]

- *Centralized vs. distributed*
 - **centralized** environment: a single monolithic system
 - all agents have access to the same structure
 - **distributed** environment: as a set of *cells* or *places* assembled in a network
 - the state of a place depends on the surrounding places
 - the perception of agents is related to one or multiple places
 - support for agent mobility, moving from place to place
- *Generalized vs specialized*
 - a generalized model is independent of the kind of actions that can be performed by agents
 - a specialized model is characterized by a well-defined set of actions

COMMUNICATIVE & SITUATED MAS

- Purely communicative MAS
 - agents can only communicate by message transfer
- Purely *situated* MAS
 - agents can only act in the environment
- Combination of communicating and situated MAS

AN ABSTRACT MODEL

- Classic model adopted in AI
 - modelling agents interacting with their environments as sequences of interleaved agent actions and environment states as responses
- System definition

$$\langle Ag, Env \rangle$$

- pair containing an agent and an environment

AGENT DEFINITION

- Agent definition

$$Ag : \mathcal{R}^E \rightarrow Ac$$

- function mapping runs ending with environment state to actions

- Agent runs \mathcal{R}

$$r : e_0 \xrightarrow{\alpha_0} e_1 \xrightarrow{\alpha_1} e_2 \xrightarrow{\alpha_2} e_3 \xrightarrow{\alpha_3} \dots \xrightarrow{\alpha_{u-1}} e_u.$$

- sequences of interleaved environment states and agent actions
 - $E = \{e, e', \dots\}$ is the finite set of the environment states
 - $Act = \{\alpha, \alpha', \dots\}$ is the finite set of agent actions

ENVIRONMENT DEFINITION

- Environment definition

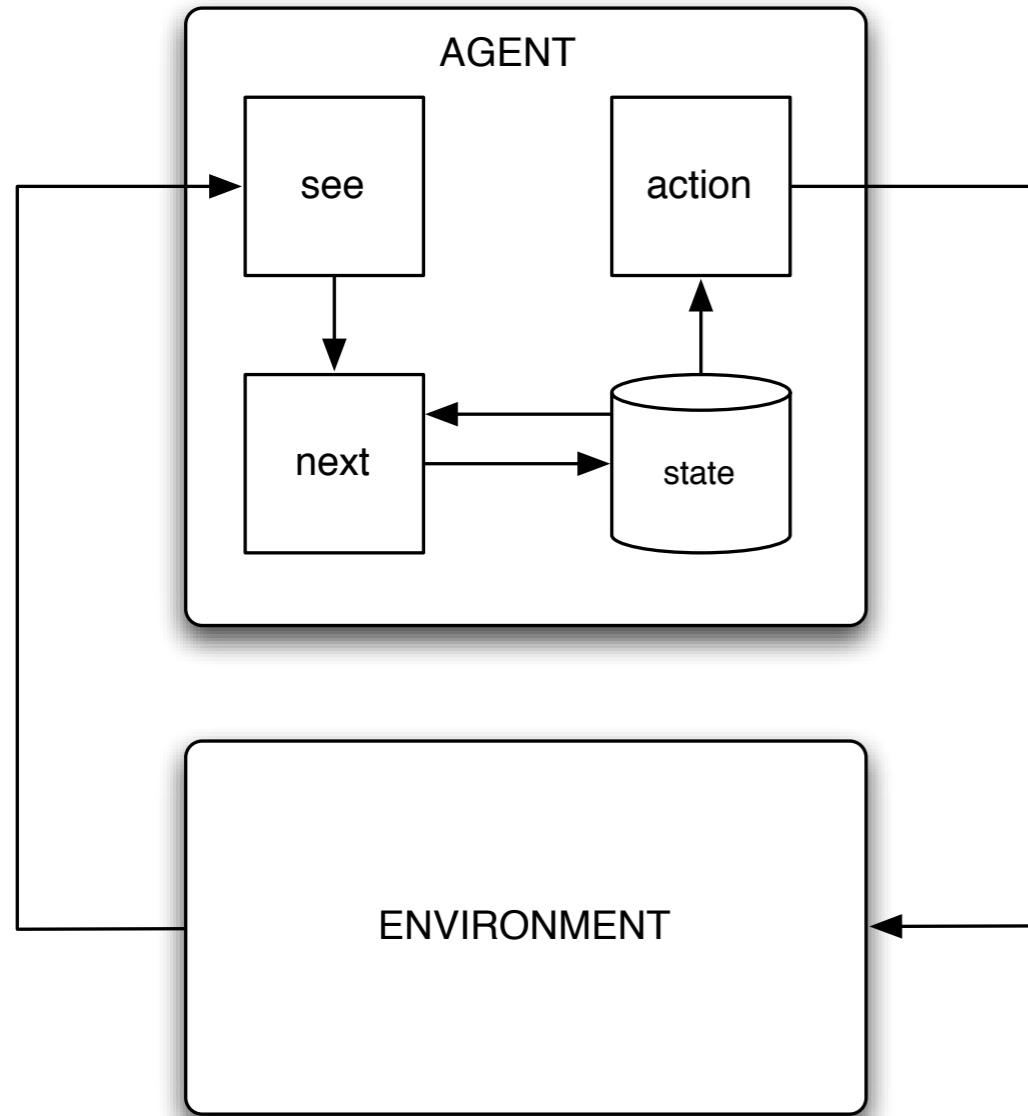
$$Env = \langle E, e_0, \tau \rangle$$

- E is the set of environment state
- $e_0 \in E$ is an initial state
- τ is the state transformer function

$$\tau : \mathcal{R}^{Ac} \rightarrow 2^E$$

- mapping a run to a set of possible environment states
 - those that could result in performing the action
- modelling the effect of agent's actions on the environment

ACTION AND PERCEPTION IN AGENT ABSTRACT ARCHITECTURE



- Perception function **see**
 - representing the ability to obtain information from its environment
$$see : E \rightarrow Per$$
- Action-selection function **action**
 - mapping from internal states to actions
$$action : I \rightarrow Ac$$
- Update state **next** function
 - mapping an internal state and a percept to an internal state
$$next : I \times Per \rightarrow I$$

REMARKS ON THE ACTION MODEL

- An action is defined as a transition of the environment state
 - as an operator whose execution produces a new state
 - from an observational point of view, the result of the behavior of an agent -its action- is directly modelled by modifying the environmental state variables
- Not fully adequate for modelling MAS in AOSE and MAS programming
 - *several agents* are acting *concurrently* on a shared environment
 - concurrent actions

RICHER ACTION MODELS

- *Influence & reactions* [Ferber and Müller, 1996]
 - distinguishing between influences and reactions to influences
 - influences come from inside the agents and are attempts to modify the course of events in the world
 - reactions, which result in state changes, are produced by the environment by combining influences of all agents, given the local state of the environment and the laws of the world
 - clear distinction between the products of the agents' behavior and the reaction of the environment
 - handling simultaneous activity in the MAS

ENVIRONMENT & REACTIVE AGENTS

- *Reactive agents*
 - no symbolic representations, no reasoning or symbolic manipulations in decision-making
- Strongly-based on the notion of environment
 - “reactive” ~ “**situated**”
 - reactive agents are typically situated in some environment rather being *disembodied*
 - agent “intelligent” behaviour is seen as innately linked to the environment agents occupy
 - the behaviour is a product of the **interaction** the agents maintain with their environment

ENVIRONMENT & INTELLIGENT AGENTS

- *Intelligent agents* [Wooldridge and Jennings, 1995]
 - reactivity
 - intelligent agents are able to **perceive the environment** and **respond in a timely fashion** to changes that occur in its order to satisfy their design objectives
 - pro-activeness
 - intelligent agents are able to exhibit *goal-directed behaviour* by taking the initiative in order to satisfy their design objectives
 - goals are defined as **state of affairs of the environment** to bring about
 - social ability
 - intelligent agents are capable of interacting with other agents in order to satisfy their design objectives
 - role of **mediated interaction**

TASK ENVIRONMENT

- The notion of *task environment* is used in literature to specify the property of the environment where the agent is situated and *the criteria by which an agent will be judged to have failed or succeeded in its task (goal)*
- Formally a task environment is defined by a pair $\langle Env, \Psi \rangle$ where:
 - Env is an environment
 - $\Psi : \mathcal{R} \rightarrow \{\text{true}, \text{false}\}$ is a *predicate* over runs
- Ψ specify the success/failure criteria.
- An agent Ag *succeeds* in task environment $\langle Env, \Psi \rangle$ if every run of Ag in Env satisfies Ψ , i.e. if $\Psi(r), \forall r \in \mathcal{R}(Ag, Env)$

ACHIEVEMENT TASKS IN THE ENVIRONMENT

- **Achievement tasks**

- “*achieve state of affairs X*”
 - specified in terms of a number of *goal states*, which typically correspond to some *environment state*
 - the agent goal is to bring about one of these goal states

- $\langle Env, \Psi \rangle$ specifies an achievement task if and only if:
 $\exists \mathcal{G} \subseteq E$ such that $\forall r \in \mathcal{R}(Ag, Env) \rightarrow \Psi(r)$, and $\exists e \in \mathcal{G}$ such that $e \in r$
 - the \mathcal{G} of an achievement task environment defines the *goal states* of the task
 - $\langle Env, \mathcal{G} \rangle$ denotes an *achievement task environment*, with goal states \mathcal{G} and environment Env

MAINTENANCE TASKS IN THE ENVIRONMENT

- **Maintenance tasks**

- “*maintain a state of affairs Y*”
 - specified in terms of environment states that the agent must *avoid*

- $\langle Env, \Psi \rangle$ specifies a maintenance task if and only if:

$\exists \mathcal{B} \subseteq E$ such that $\Psi(r)$, and $\forall e \in \mathcal{B}$ we have $e \notin r, \forall r \in \mathcal{R}(Ag, Env)$

- the \mathcal{B} of a maintenance task environment defines the *failure set* of the task
- $\langle Env, \mathcal{B} \rangle$ denotes a *maintenance task environment*, with environment Env and failure set \mathcal{B}

GAME VIEW

- Achievement tasks as *the agent playing a game against the environment*
 - environment and agents begin in some state
 - the agent takes a turn by executing an action
 - the env. responds with some state, and so on
 - the agent wins if it can force the env. into one of the goal state
- Maintenance tasks as games in which the agents win if it manages to avoid all the states in the failure set
 - environment as the opponent trying to force the agent into failure set

REMARKS

- Traditional AI perspective: Environment is reduced to a discrete space to explore in problem solving
 - the “task environment” is a search space of choices and outcomes (i.e., computer playing chess)
 - problem solving is an activity that take place wholly within an agent mind (problem solver)
- But the world involves complex interactions, between heterogenous entities
 - actions may have uncertain outcomes, uncertain knowledge

BOUNDED RATIONALITY

- Real, complex environments make the assumption of agents unbounded rationality *implausible*
 - Bounded Rationality vs. Rational Choice in economics
 - agents can plausibly exploit a subset of their resources to solve in a timely fashion a given problem [Simon, 1947]
 - computational resources needed to explore the search space are finite
 - Frame Problem [McCarthy & Hayes, 1969]
 - need to explicitly specify which conditions are not affected by any given action
 - Representational Content in Machines [Bickhard, 1993]
 - how translate raw percepts to formal, computable symbols
- ▶ BDI Agents and Practical Reasoning

ACTIONS AND PERCEPTIONS IN BDI AGENTS

- BDI agent model/architecture
 - agent programs specified in terms of mental attitudes (beliefs, desires, intentions)
 - based on *practical reasoning*
 - deliberation
 - deciding *what* states of affairs to achieve / maintain
 - means-ends reasoning
 - *how* to achieve / maintain the states of affairs
- Action and perception stage as part of the agent execution cycle

BDI AGENT EXECUTION CYCLE

```
B ← B0; /* B0 are initial beliefs */  
I ← I0; /* I0 are initial intentions */  
while true do  
    get next percept  $\rho$  through see(...) function;  
    B ← brf(B,  $\rho$ );  
    D ← options(B, I);  
    I ← filter(B, D, I);  
     $\pi$  ← plan(B, I, Ac);  
    while not (empty( $\pi$ ) or succeeded(I, B) or impossible(I, B)) do  
         $\alpha$  ← hd( $\pi$ );  
        execute( $\alpha$ );  
         $\pi$  ← tail( $\pi$ );  
        get next percept  $\rho$  through see(...) function;  
        B ← brf(B,  $\rho$ );  
        if reconsider(I, B) then  
            D ← options(B, I);  
            I ← filter(B, D, I);  
        end-if  
        if not sound( $\pi$ , I, B) then  
             $\pi$  ← plan(B, I, Ac);  
        end-if  
    end-while  
end-while
```