

INTELLIGENT AGENTS¹

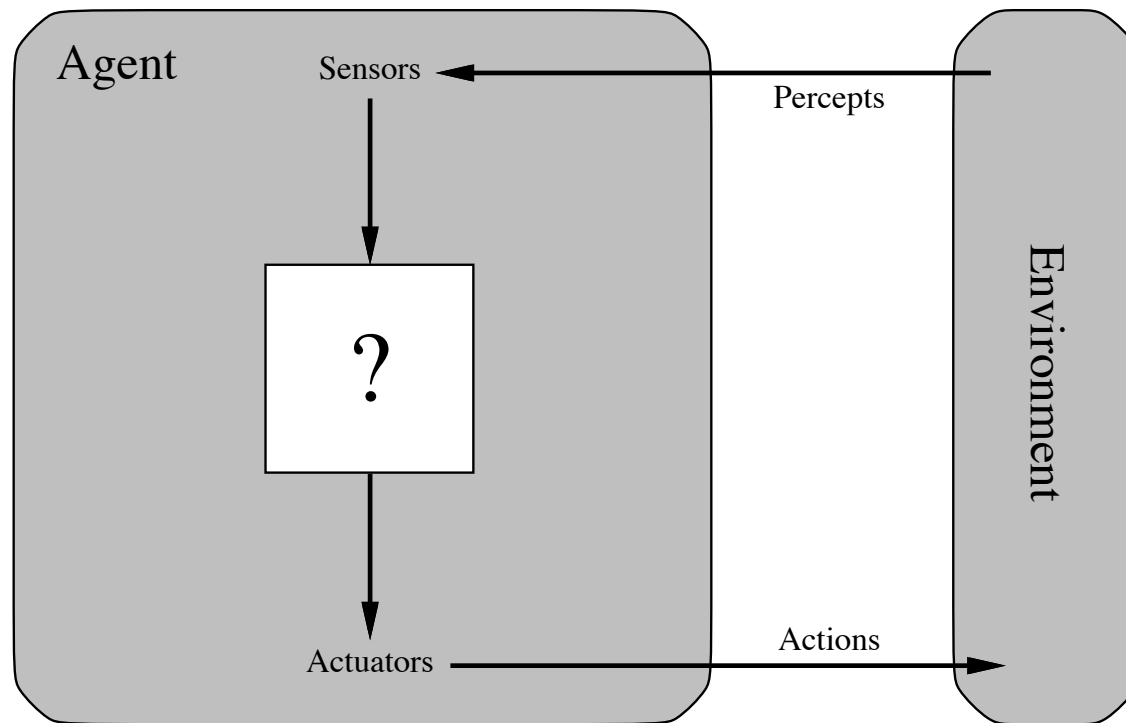
LECTURE 2

¹The slides have been prepared using the textbook material available on the web, and the slides of the previous editions of the course by Prof. Luigia Carlucci Aiello

Summary

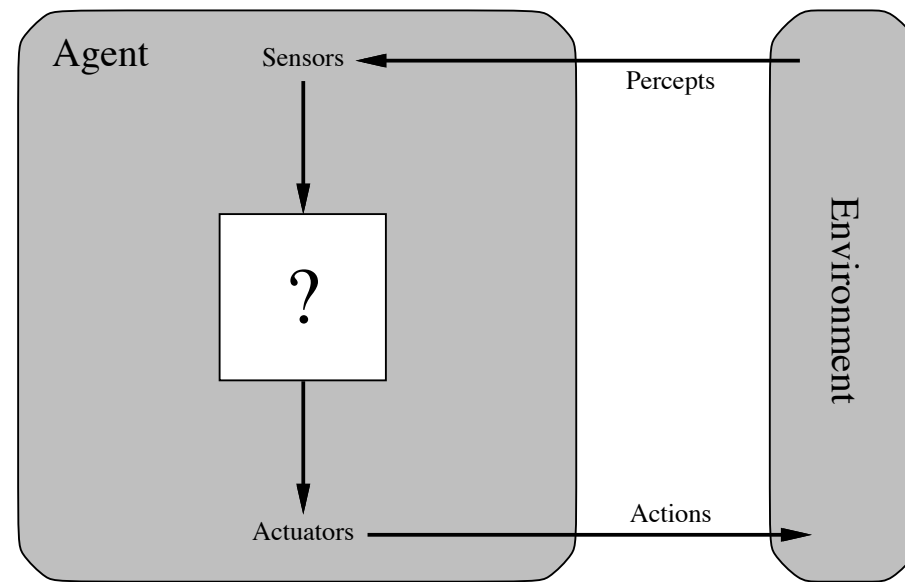
- ◇ Agents and Environments
- ◇ Functions and Programs for Agents
- ◇ Environment specification
- ◇ Environment types
- ◇ Agent types

Agent and Environment



Agents include humans, robots, softbots, thermostats, etc.

Agent architecture

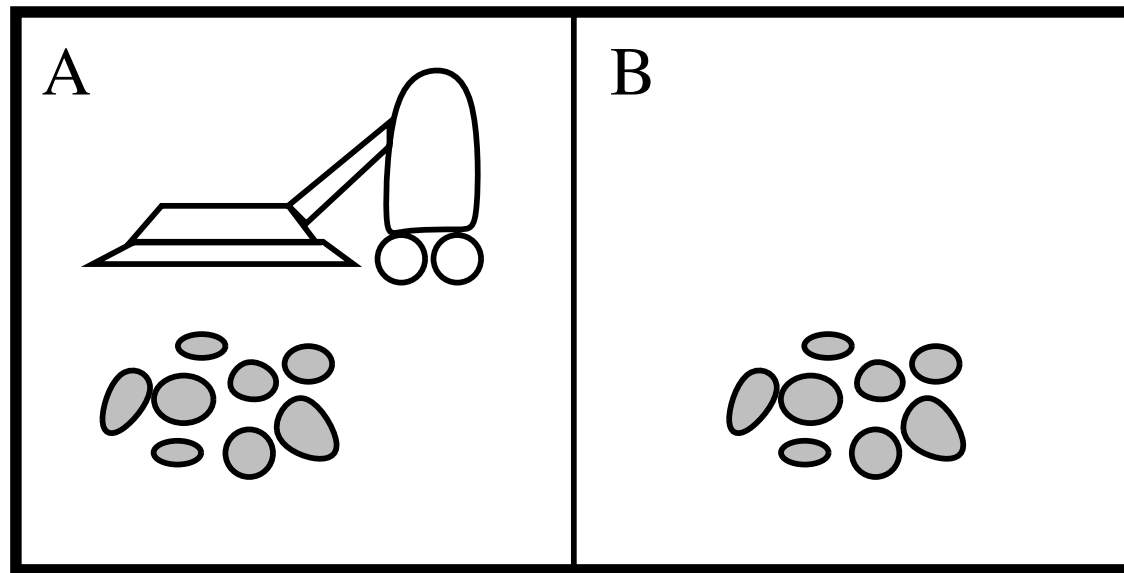


The **agent function** maps from percept histories to actions:

$$f : \mathcal{P}^* \rightarrow \mathcal{A}$$

The **agent program** runs on a physical **architecture** to produce f

Vacuum-cleaner world



Percepts: location and contents, e.g., $[A, \textit{Dirty}]$

Actions: *Left*, *Right*, *Suck*, *NoOp*

Rationality

Fixed **performance measure** evaluates the **environment sequence**

- one point per square cleaned up in time T ?
- one point per clean square per time step, minus one per move?
- penalize for $> k$ dirty squares?

A **rational agent** chooses whichever action maximizes the **expected** value of the performance measure **given the percept sequence to date**

Rational \neq omniscient

Rational \neq clairvoyant

Rational \neq successful

Rational \Rightarrow learning, exploration (autonomy)

Functions and agent programs

Agent specification: agent function which maps sequences of percepts into actions

Idealistic implementation: table; Pratical implementation: a program, that may also keep track of the sequence of input percepts

```
function SKELETON-AGENT(percept) returns action
  static: memory, agent's memory of the world
  memory ← UPDATE-MEMORY(memory, percept)
  action ← CHOOSE-BEST-ACTION(memory)
  memory ← UPDATE-MEMORY(memory, action)
  return action
```

PEAS

To design a rational agent, we must specify the **task environment**

Consider, e.g., the task of designing an automated taxi:

Performance measure?? safety, destination, profits, legality, comfort, ...

Environment?? US streets/freeways, traffic, pedestrians, weather, ...

Actuators?? steering, accelerator, brake, horn, speaker/display, ...

Sensors?? video, accelerometers, gauges, engine sensors, keyboard, GPS, ...

Soccer Robot

Performance??

Environment??

Actions??

Sensors??

Environment's Features

- ◇ Observable (Partially)
- ◇ Deterministic (non-deterministic, stochastic)
- ◇ Episodic (Sequential)
- ◇ Static (Dynamic, Semidynamic)
- ◇ Discrete (Continuous)
- ◇ Single Agent (Multi)

The environment influences the agent design

Environment types

	Soli	Backgam	Image	Ishop	Taxi
<u>Observable??</u>	Yes	Yes	Yes	No	No
<u>Deterministic??</u>	Yes	No	Yes	Partly	No
<u>Episodic??</u>	No	No	Yes	No	No
<u>Static??</u>	Yes	Semi	Semi	Semi	No
<u>Discrete??</u>	Yes	Yes	No	Yes	No
<u>Single-agent??</u>	Yes	No	Yes	Yes (*)	No

The environment type largely determines the agent design

The real world is (of course) partially observable, stochastic, sequential, dynamic, continuous, multi-agent

Vacuum cleaner world

Performance

- +100 for every dirt sucked
- -1 for every action
- -1000 to turn off not at home position

Environment

- square grid, with walls and obstacles
- creation and distribution of dirt, bag
- motion actions: move the agent when no obstacles
- sucking action: puts dirt in the bag

Sensors (<bump> <dirt> <home>)

Actions turnoff forward suck (turnleft) (turnright)

Observable? Deterministic? Episodic? Static? Discrete?

Environment Simulation

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*] \leftarrow GET-PERCEPT(*agent*, *state*)

end

for each *agent* **in** *agents* **do**

 ACTION[*agent*] \leftarrow PROGRAM[*agent*](PERCEPT[*agent*])

end

state \leftarrow UPDATE-FN(*actions*, *agents*, *state*)

until *termination*(*state*)

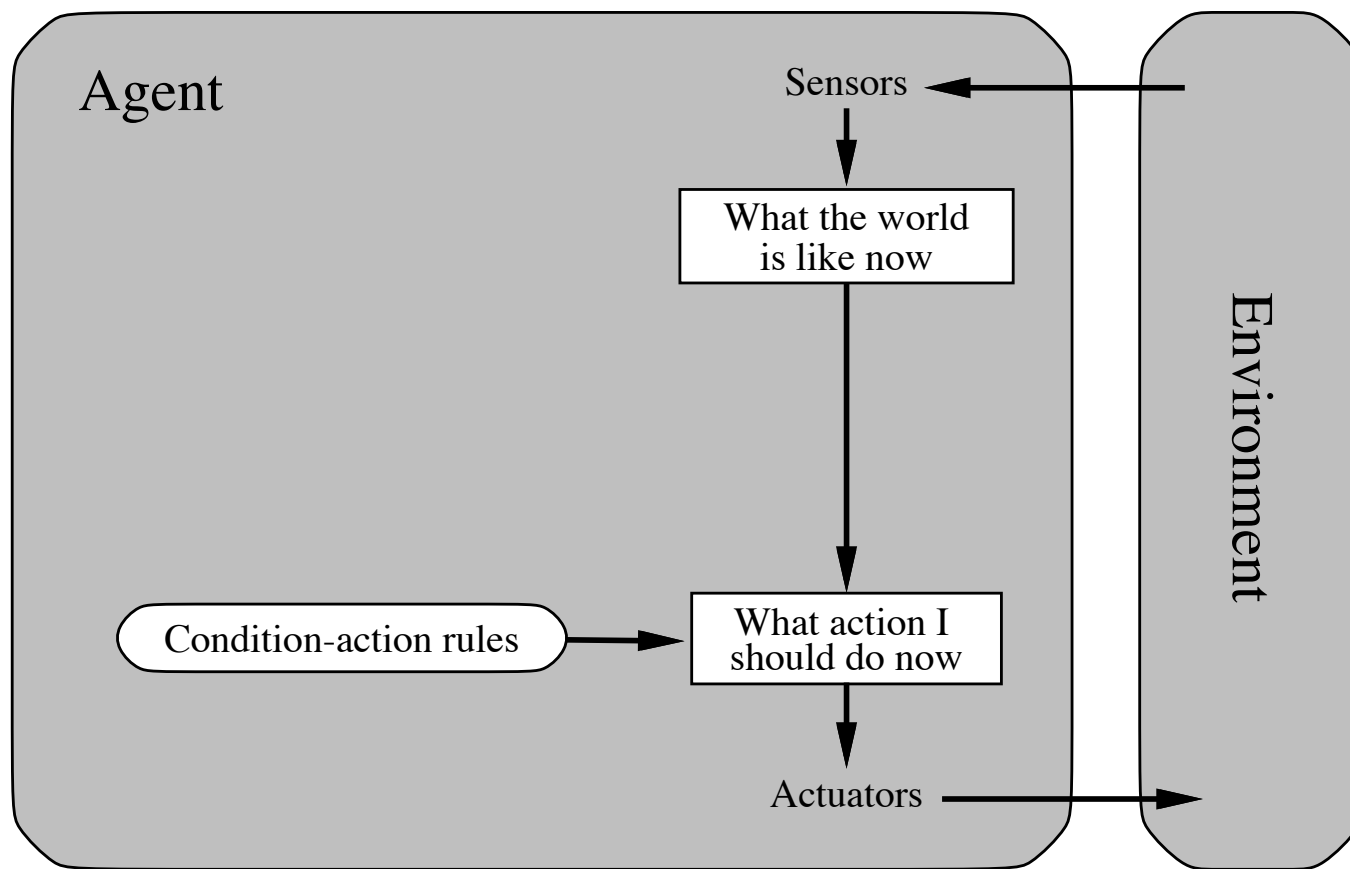
Agent types

Four basic types in order of increasing generality:

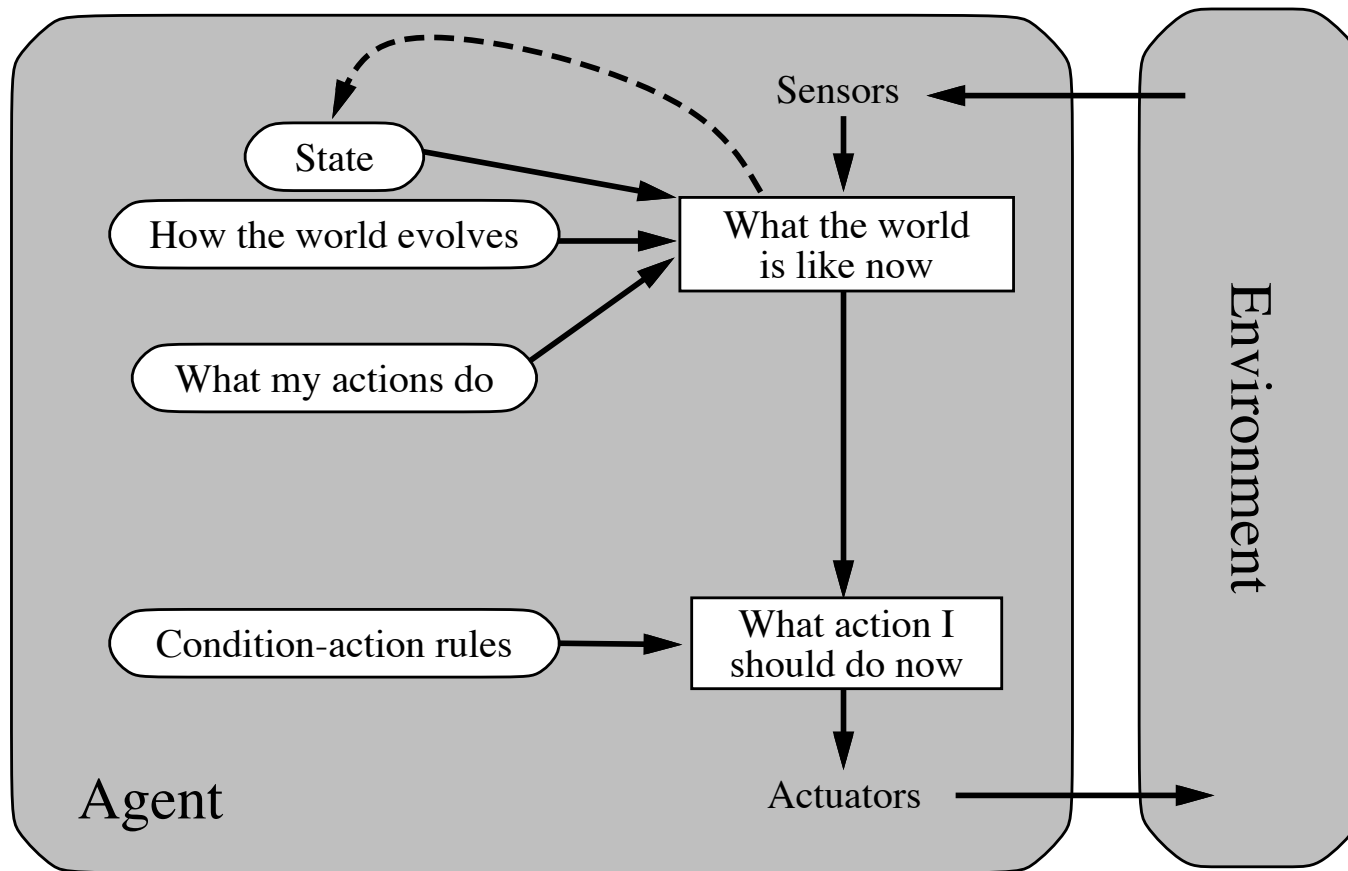
- simple reflex agents
- reflex agents with state
- goal-based agents
- utility-based agents

All these can be turned into learning agents

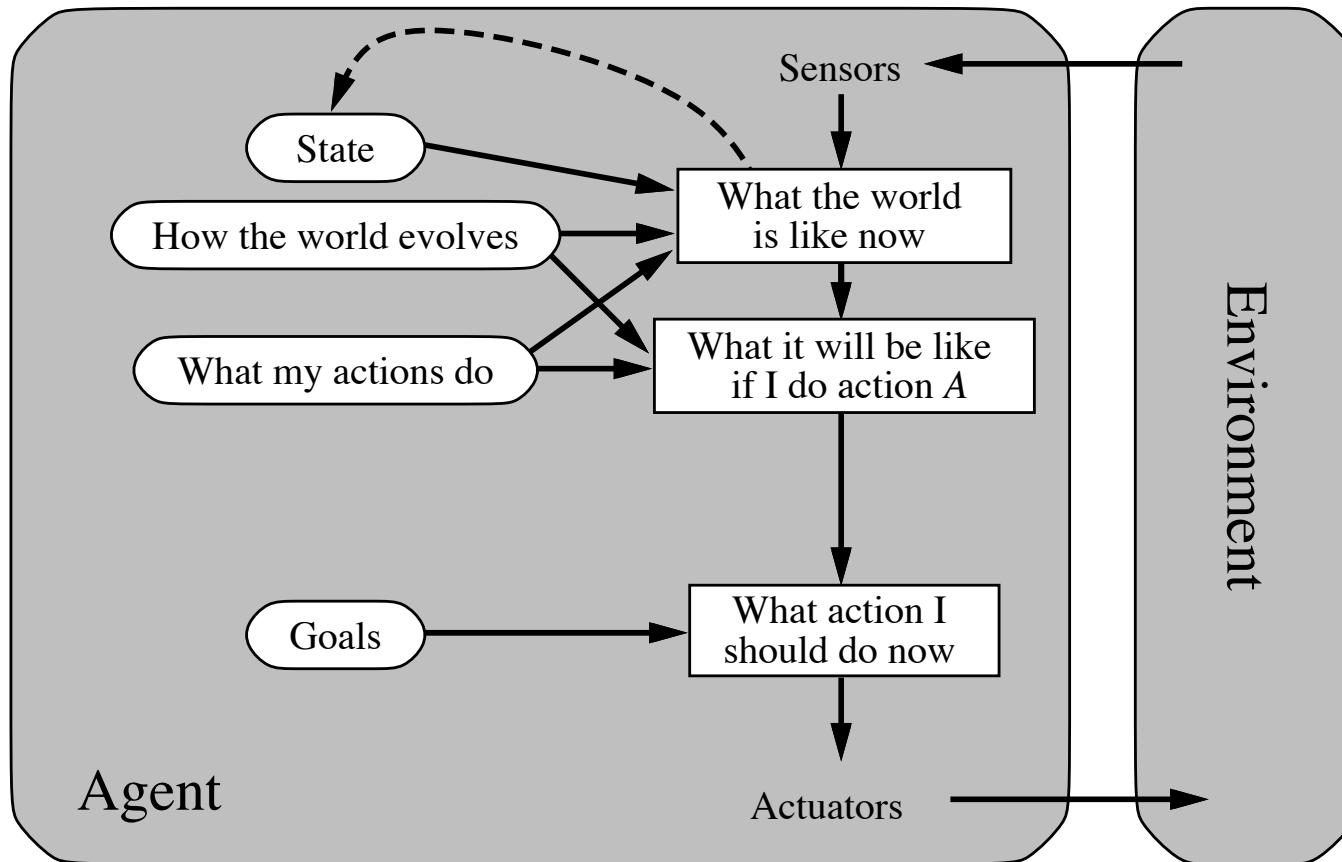
Simple reflex agents



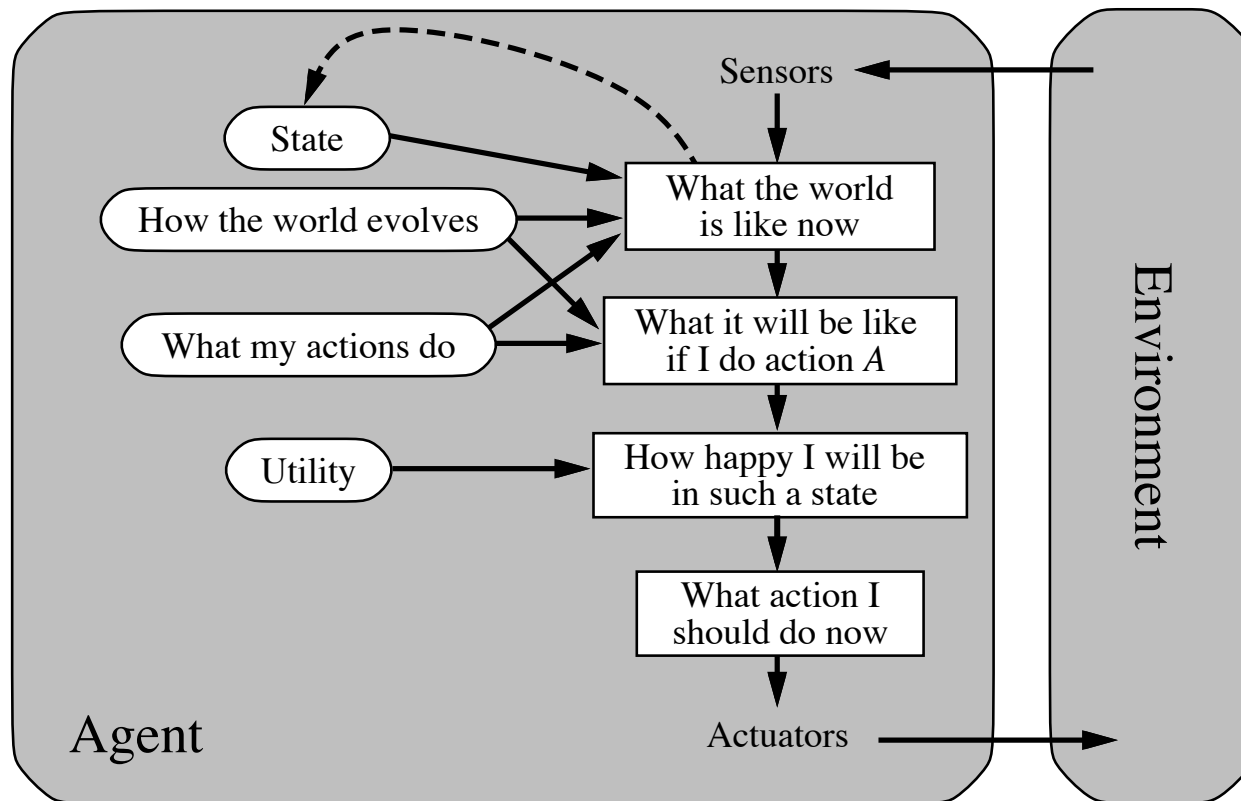
Reflex agents with state



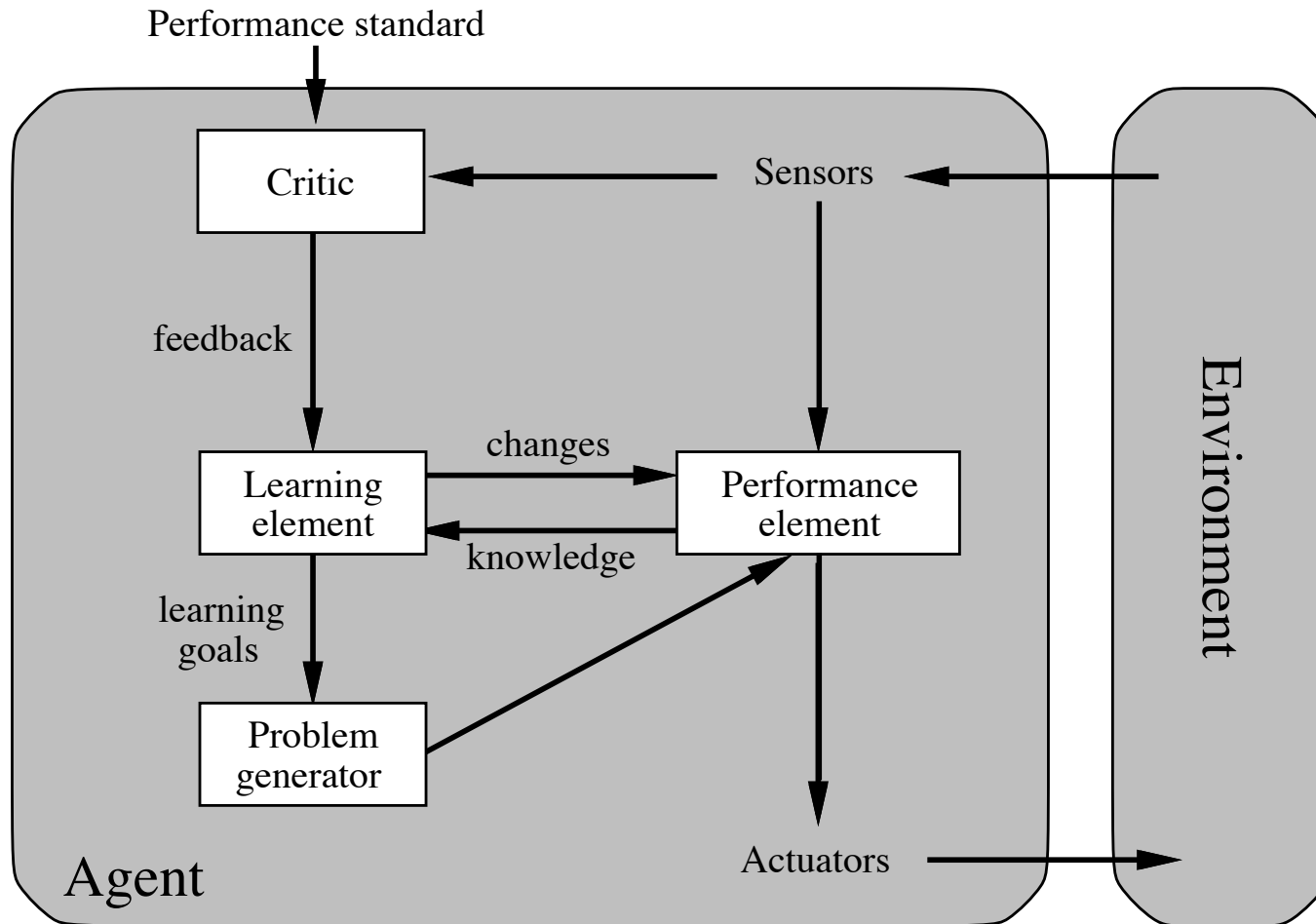
Goal-based agents



Utility-based agents



Learning agents



Summarizing

AI agents have the following features:

- ◇ Perception
- ◇ Reasoning
- ◇ Action

Note:

- ◇ applicable robots and softbot
- ◇ integration of all the above
- ◇ a chess player robot should perceive the board, the moves . . .