

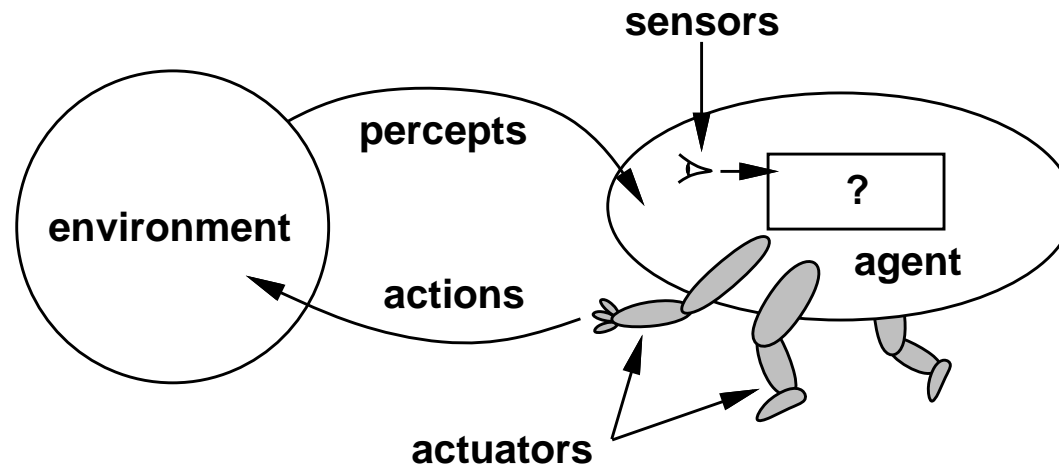
INTELLIGENT AGENTS

CHAPTER 2, SECTIONS 1–4

Outline

- ◇ Agents and environments
- ◇ Rationality
- ◇ PEAS (Performance measure, Environment, Actuators, Sensors)
- ◇ Environment types
- ◇ Agent types

Agents and environments



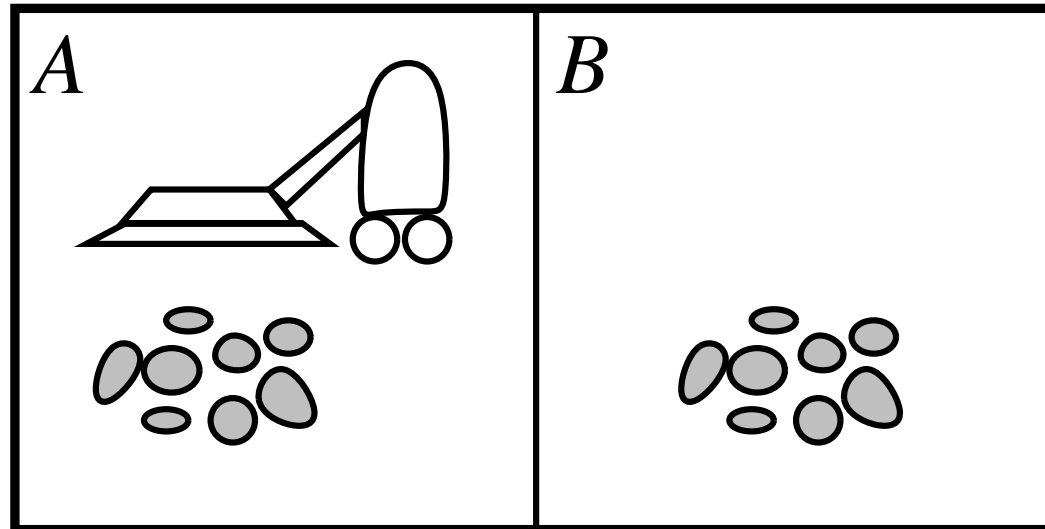
Agents include humans, robots, softbots, thermostats, etc.

The agent function maps from percept histories to actions:

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

The agent program runs on the physical architecture to produce f

Vacuum-cleaner world



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

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

A vacuum-cleaner agent

A simple agent function is:

If the current square is dirty, then suck;
otherwise, move to the other square.

... or as pseudo-code:

```
function REFLEX-VACUUM-AGENT( [location,status] ) returns an action
  if status = Dirty then return Suck
  else if location = A then return Right
  else if location = B then return Left
```

How do we know if this is a **good** agent function?

What is the **best** function?

Is there one?

Who decides this?

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 any action that

- maximizes the **expected** value of the performance measure
- **given the percept sequence to date**

Rational \neq omniscient

- percepts may not supply all relevant information

Rational \neq clairvoyant

- action outcomes may not be as expected

Hence, rational \neq successful

Rational \Rightarrow exploration, learning, autonomy

PEAS

To design a rational agent, we must specify the **task environment**, which consists of the following four things:

Performance measure??

Environment??

Actuators??

Sensors??

Examples of agents:

- Automated taxi,
- Internet shopping agent,
- Boardgames

Automated taxi

The task environment for an automated taxi:

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

Environment?? Streets, traffic, pedestrians, weather, ...

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

Sensors?? Video, accelerometers, gauges, engine, keyboard, GPS, ...

Internet shopping agent

The task environment for an internet shopping agent:

Performance measure?? Price, quality, appropriateness, efficiency

Environment?? Current and future WWW sites, vendors, shippers

Actuators?? Display to user, follow URL, fill in form

Sensors?? HTML pages (text, graphics, scripts)

Question-answering system

The task environment for a question-answering system:

Performance measure?? User satisfaction? Known questions?

Environment?? Wikipedia, Wolfram alpha, ontologies, encyclopedia, . . .

Actuators?? Spoken/written language

Sensors?? Written/spoken input

Question-answering system: **Jeopardy!**

The task environment for a question-answering system:

Performance measure?? \$ \$ \$

Environment?? Wikipedia, Wolfram alpha, ontologies, encyclopedia, . . .

Actuators?? Spoken/written language, answer button

Sensors?? Written/spoken input

Why did IBM choose Jeopardy as its goal for a QA system?
Because of the performance measure!

Environment types

	Solitaire	Backgammon	Web shopping	Taxi
<u>Observable??</u>	Fully	Fully	Partly	Partly
<u>Deterministic??</u>	Deterministic	Stochastic	Partly	Stochastic
<u>Episodic??</u>	Sequential	Sequential	Sequential	Sequential
<u>Static??</u>	Static	Static	Semi	Dynamic
<u>Discrete??</u>	Discrete	Discrete	Discrete	Continuous
<u>Single-agent??</u>	Single	Multi	Single*	Multi

*except auctions

The environment type largely determines the agent design

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

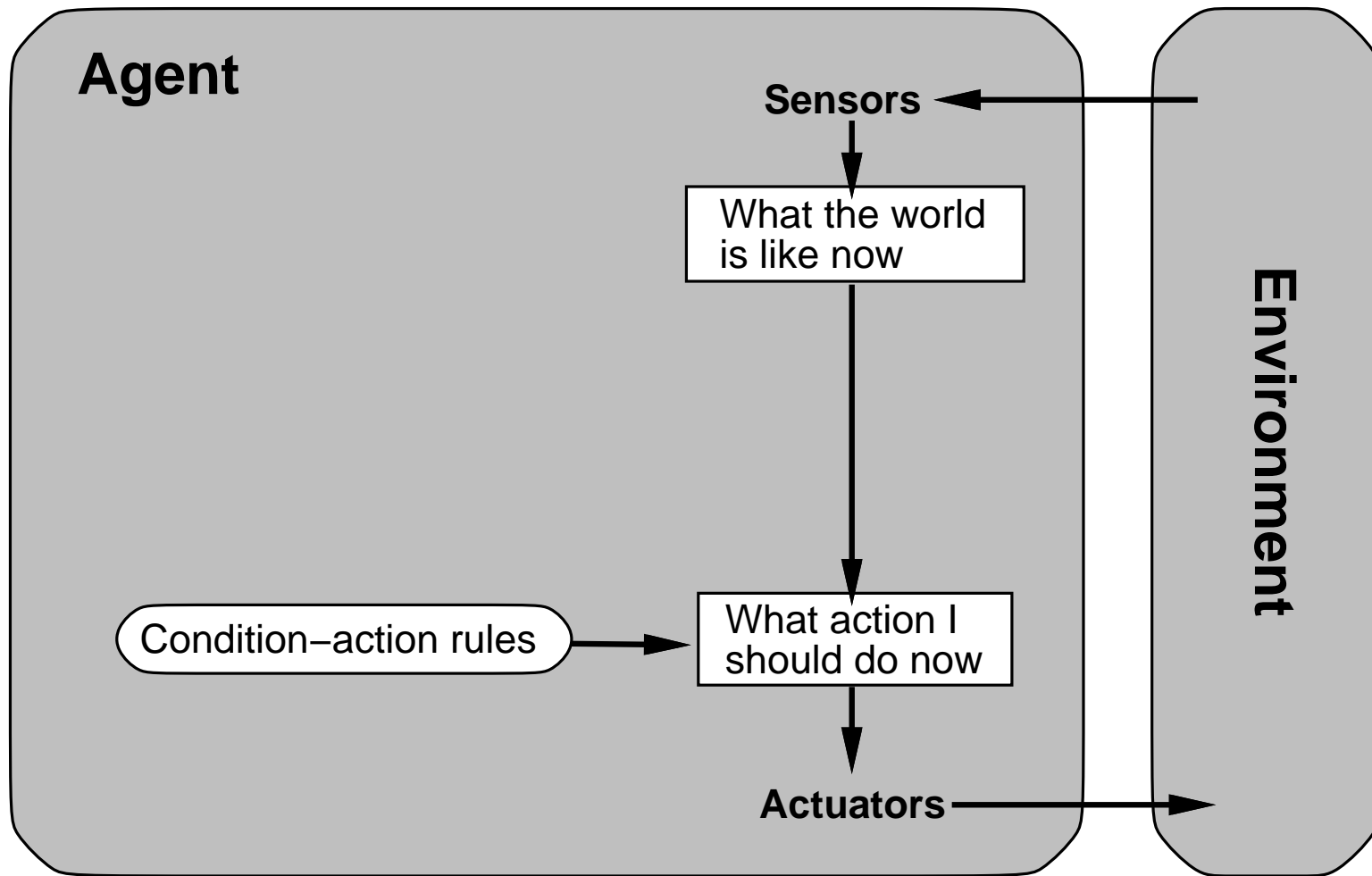
Agent types

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



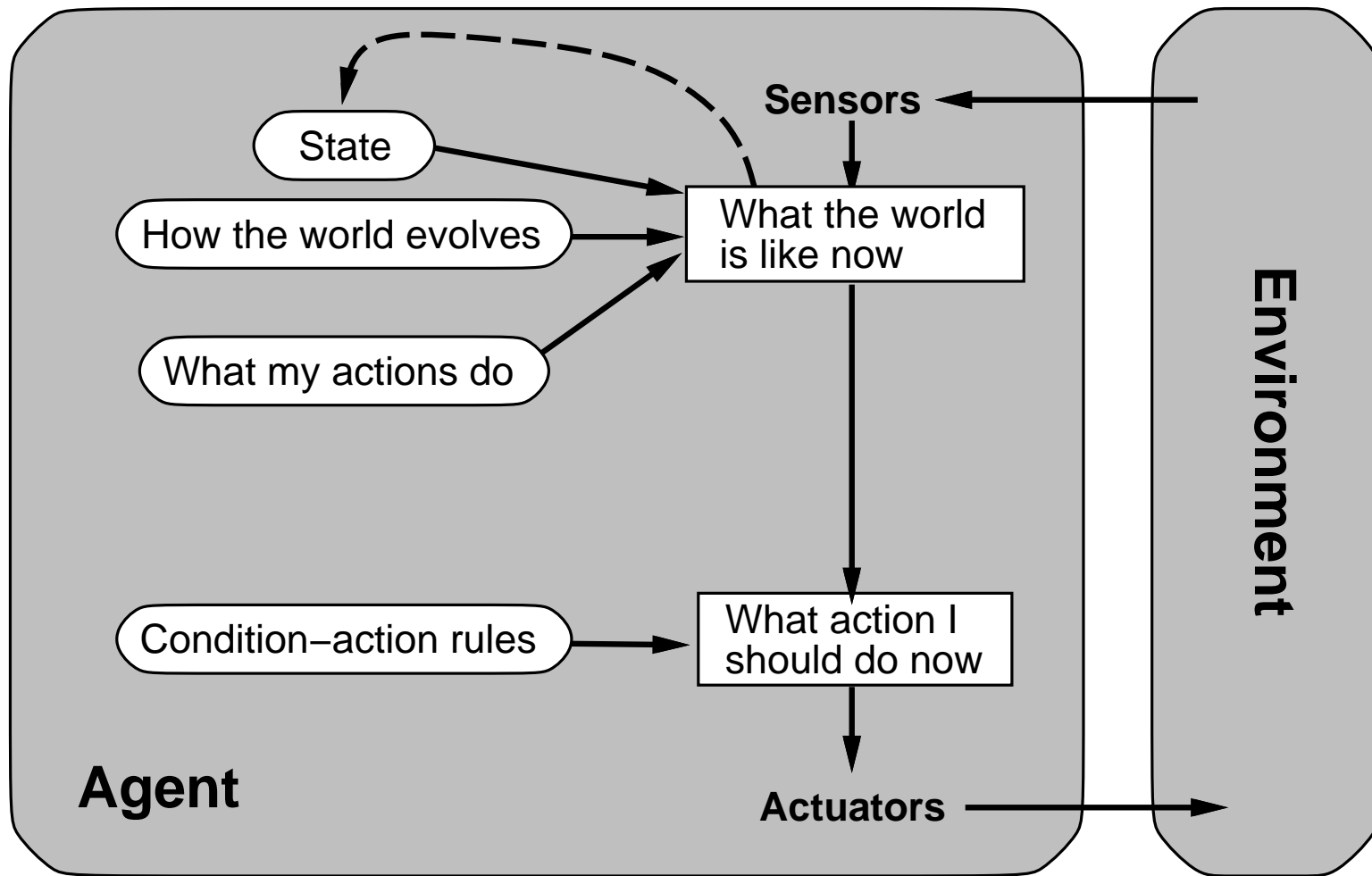
Example

function REFLEX-VACUUM-AGENT([*location, status*]) **returns** an action

if *status* = *Dirty* **then return** *Suck*
else if *location* = *A* **then return** *Right*
else if *location* = *B* **then return** *Left*

```
def reflex_vacuum_agent(sensors):  
    if sensors['status'] == 'Dirty':  
        return 'Suck'  
    elif sensors['location'] == 'A':  
        return 'Right'  
    elif sensors['location'] == 'B':  
        return 'Left'
```

Reflex agents with state

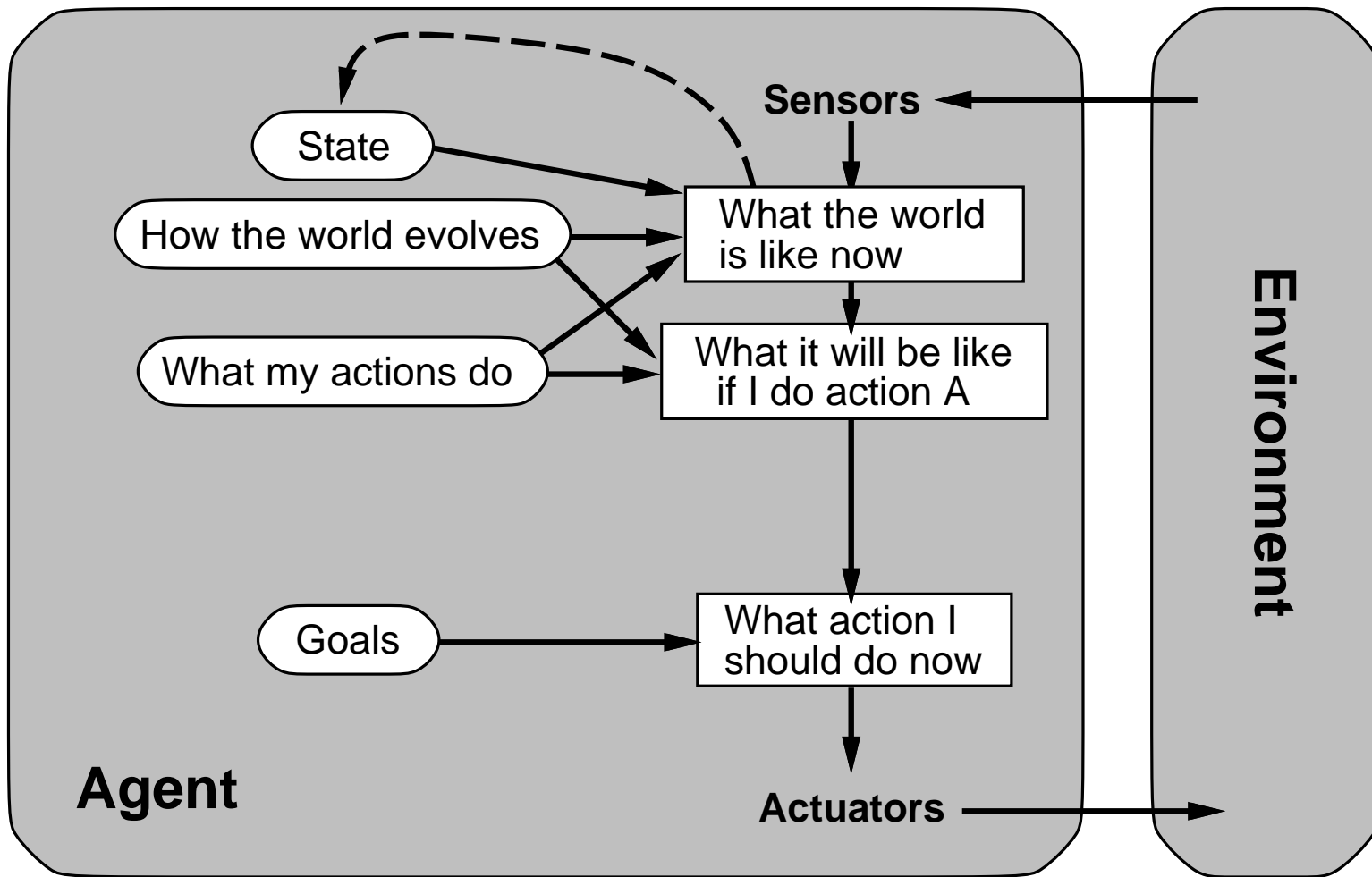


Example

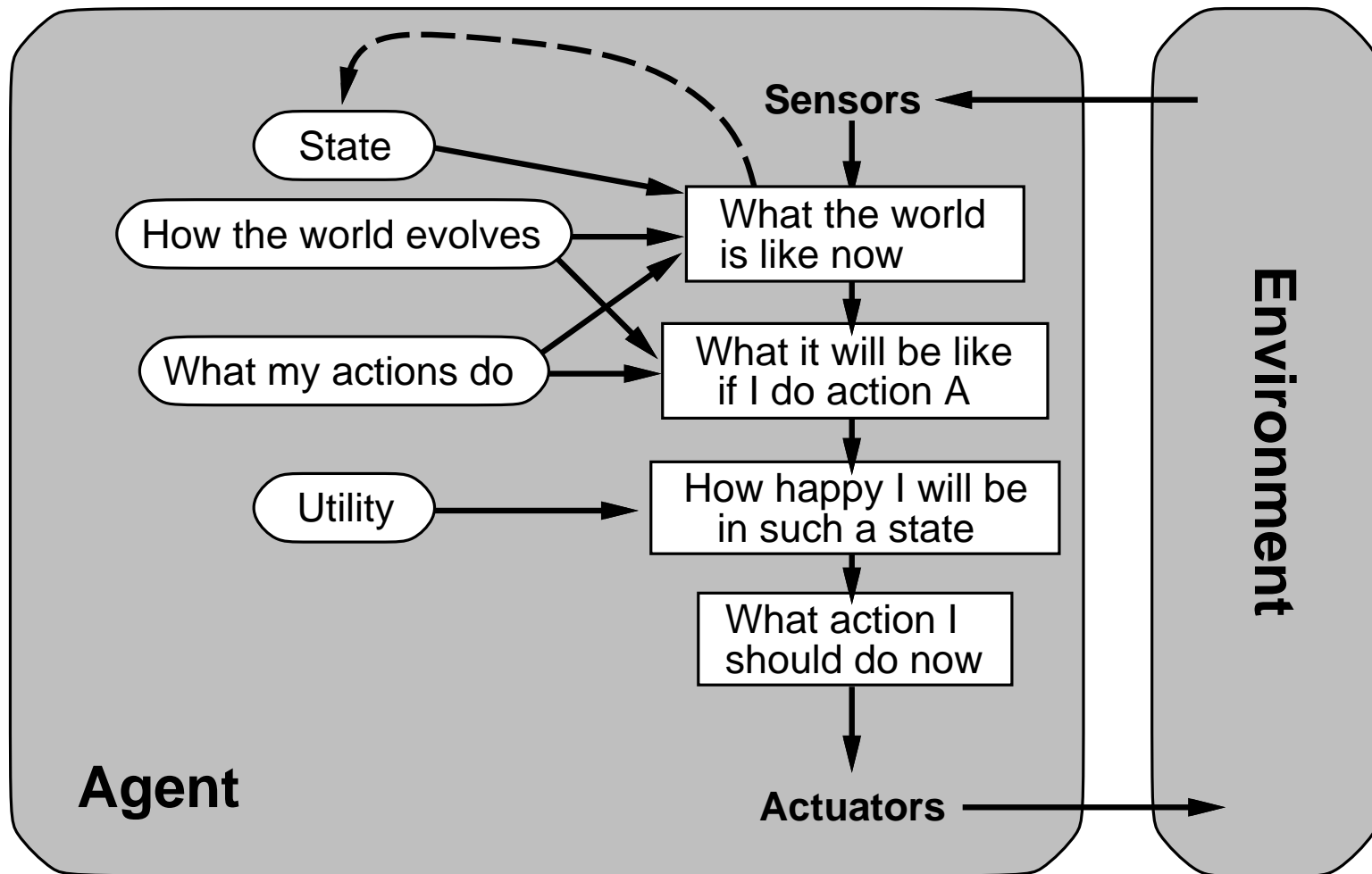
function REFLEX-VACUUM-AGENT(*[location, status]*) **returns** an action
static: *last_A, last_B*, numbers, initially ∞
if *status* = *Dirty* **then** ...

```
initial_state = {'last-A': maxint, 'last-B': maxint}
def reflex_vacuum_agent_state(sensors, state=initial_state):
    if sensors['status'] == 'Dirty':
        if sensors['location'] == 'A':
            state['last-A'] = 0
        else:
            state['last-B'] = 0
        return 'Suck'
    elif sensors['location'] == 'A' and state['last-B'] > 3:
        return 'Right'
    elif sensors['location'] == 'B' and state['last-A'] > 3:
        return 'Left'
    else:
        return 'NoOp'
```

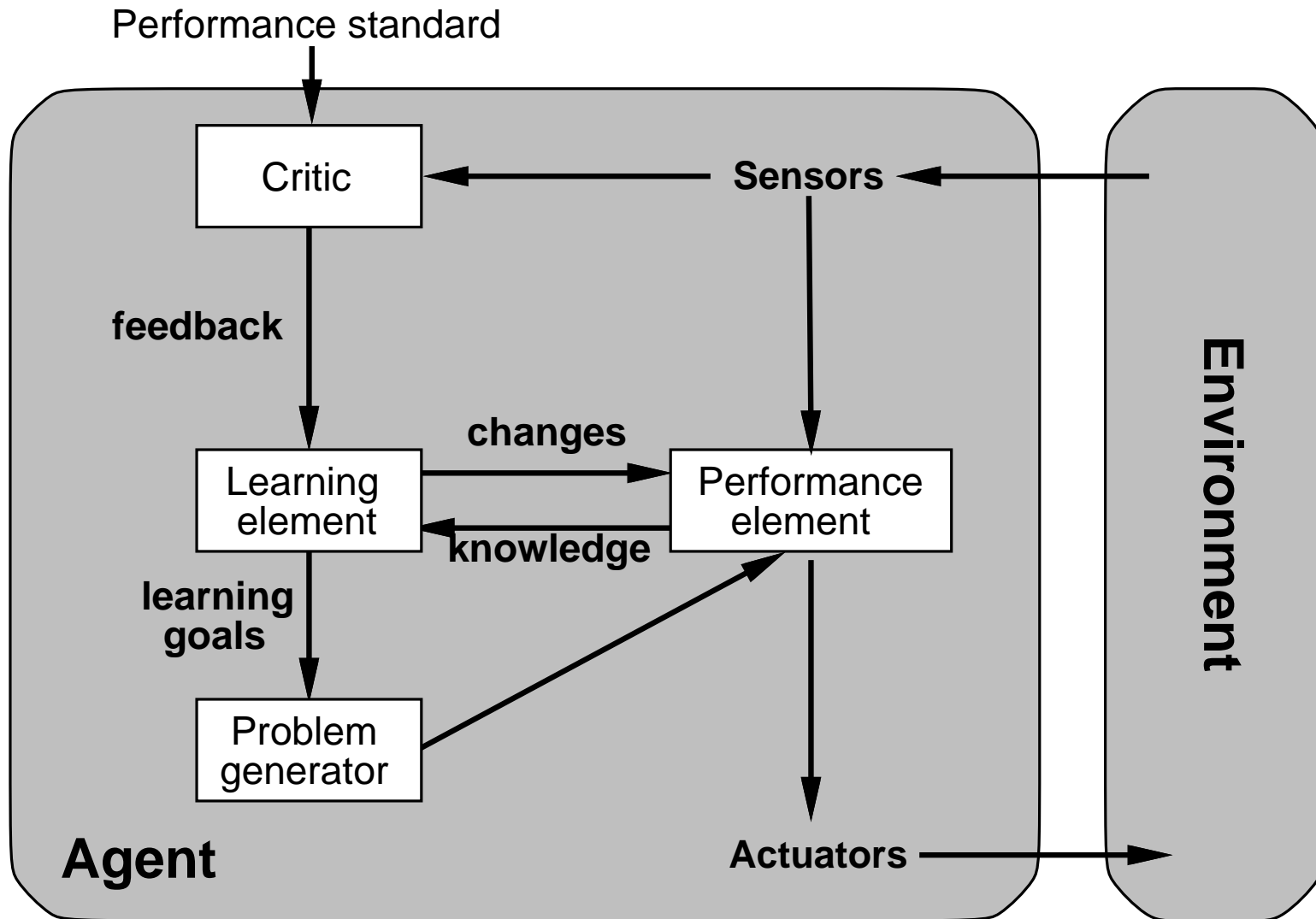
Goal-based agents



Utility-based agents



Learning agents



Summary

Agents interact with environments through actuators and sensors

The agent function describes what the agent does in all circumstances

The performance measure evaluates the environment sequence

A perfectly rational agent maximizes expected performance

Agent programs implement (some) agent functions

PEAS descriptions define task environments

Environments are categorized along several dimensions:

observable? deterministic? episodic? static? discrete? single-agent?

Several basic agent architectures exist:

reflex, reflex with state, goal-based, utility-based