

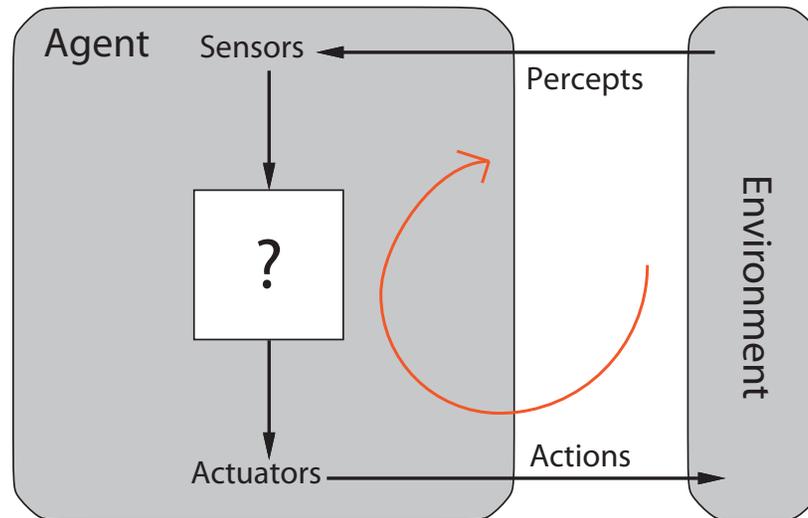
# INTELLIGENT AGENTS

## CHAPTER 2

# Outline

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

# Agents and environments



Perception-Action Cycle

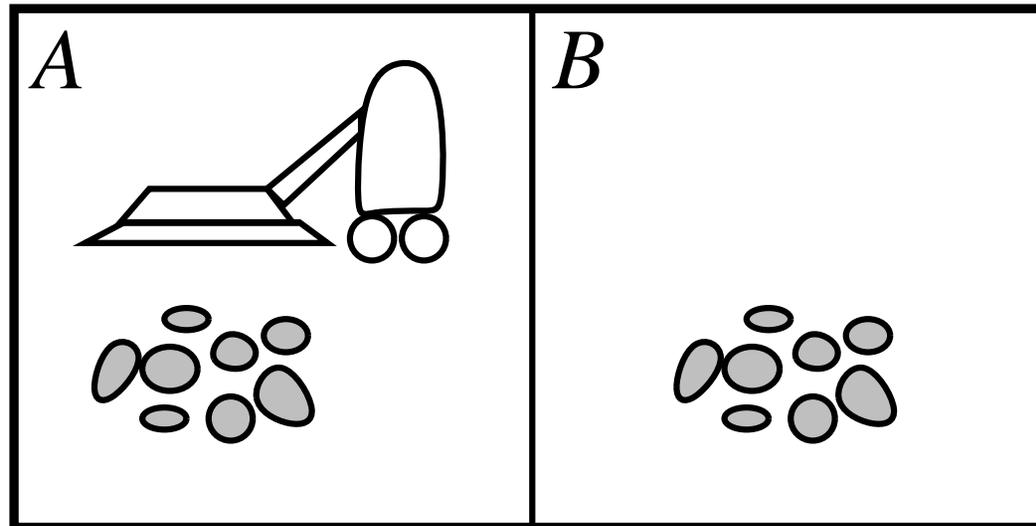
An **agent** is anything that can be viewed as perceiving its **environment** through **sensors** and acting upon that environment through **actuators**.

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  $[A, B]$  and contents  $[Dirty, Clean]$ , e.g.,  $[A, Dirty]$

How many state combinations are possible?  $2 \times 2^2 = 8$

How about when we have more advanced vacuum? vacuum  $[on/off/sleep]$  and more locations  $[1 \ 10]$ ? 3072

Actions: *Left, Right, Suck, NoOp*

## A vacuum-cleaner agent

Percept sequence	Action
$[A, \textit{Clean}]$	$\textit{Right}$
$[A, \textit{Dirty}]$	$\textit{Suck}$
$[B, \textit{Clean}]$	$\textit{Left}$
$[B, \textit{Dirty}]$	$\textit{Suck}$
$[A, \textit{Clean}], [A, \textit{Clean}]$	$\textit{Right}$
$[A, \textit{Clean}], [A, \textit{Dirty}]$	$\textit{Suck}$
$\vdots$	$\vdots$

```
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
```

What is the **right** function?

Can it be implemented in a small agent program?

# 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

- 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**

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

Performance measure??

Environment??

Actuators??

Sensors??

# 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, ...

# Internet shopping agent

Performance measure??

Environment??

Actuators??

Sensors??

# 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)

More examples in AIMA Fig. 2.5

# Properties of task environments

- ◇ Fully observable vs partially observable.
- ◇ Deterministic vs stochastic
- ◇ Episodic vs sequential
- ◇ Statics vs dynamic
- ◇ Discrete vs continuous
- ◇ Benign vs adversarial

## Environment types

	Solitaire	Backgammon	Internet shopping	Taxi
<u>Observable??</u>				
<u>Deterministic??</u>				
<u>Episodic??</u>				
<u>Static??</u>				
<u>Discrete??</u>				
<u>Single-agent??</u>				

## Environment types

	Solitaire	Backgammon	Internet shopping	Taxi
<u>Observable??</u>	Yes	Yes	No	No
<u>Deterministic??</u>	Yes	No	Partly	No
<u>Episodic??</u>	No	No	No	No
<u>Static??</u>	Yes	Semi	Semi	No
<u>Discrete??</u>	Yes	Yes	Yes	No
<u>Single-agent??</u>	Yes	No	Yes (except auctions)	No

**The environment type largely determines the agent design**

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

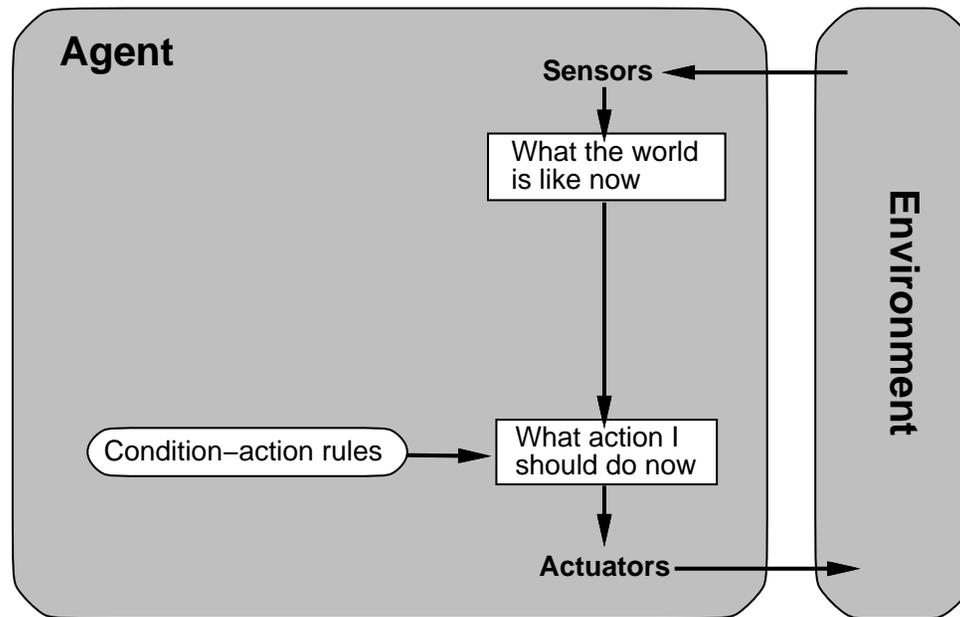
# Agent types

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

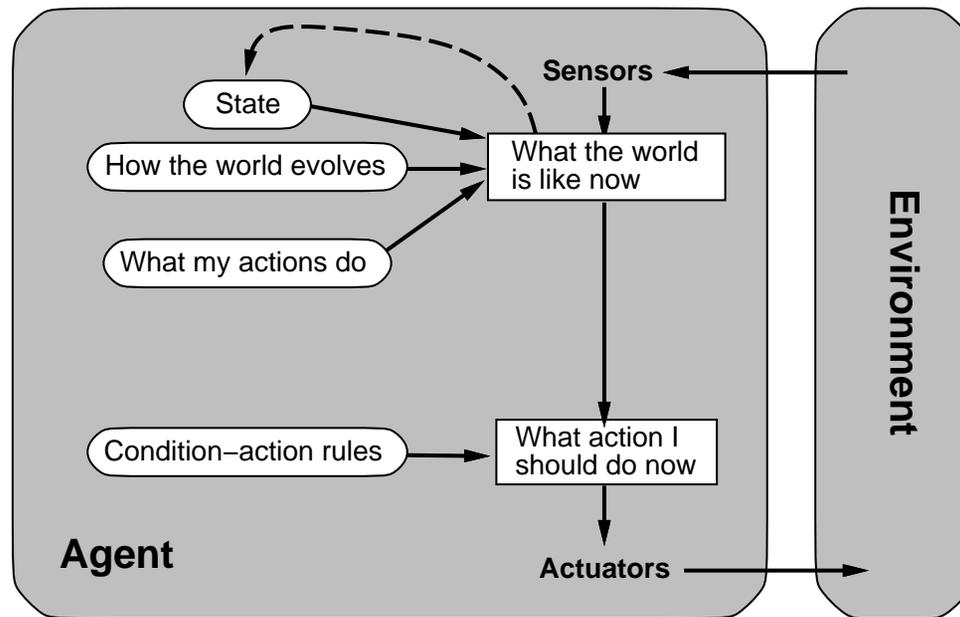
# Simple reflex agents



## 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
```

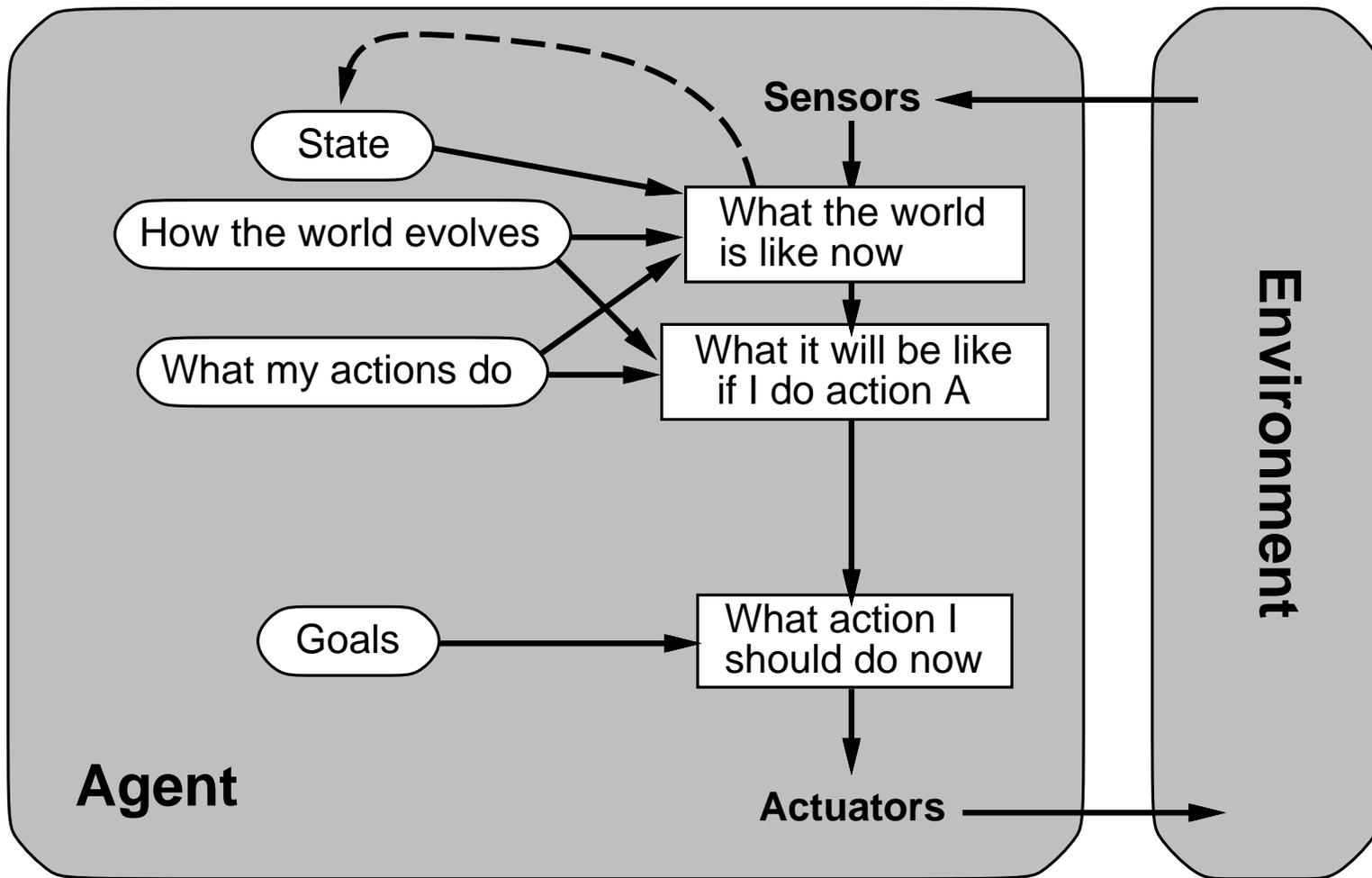
# Reflex agents with state



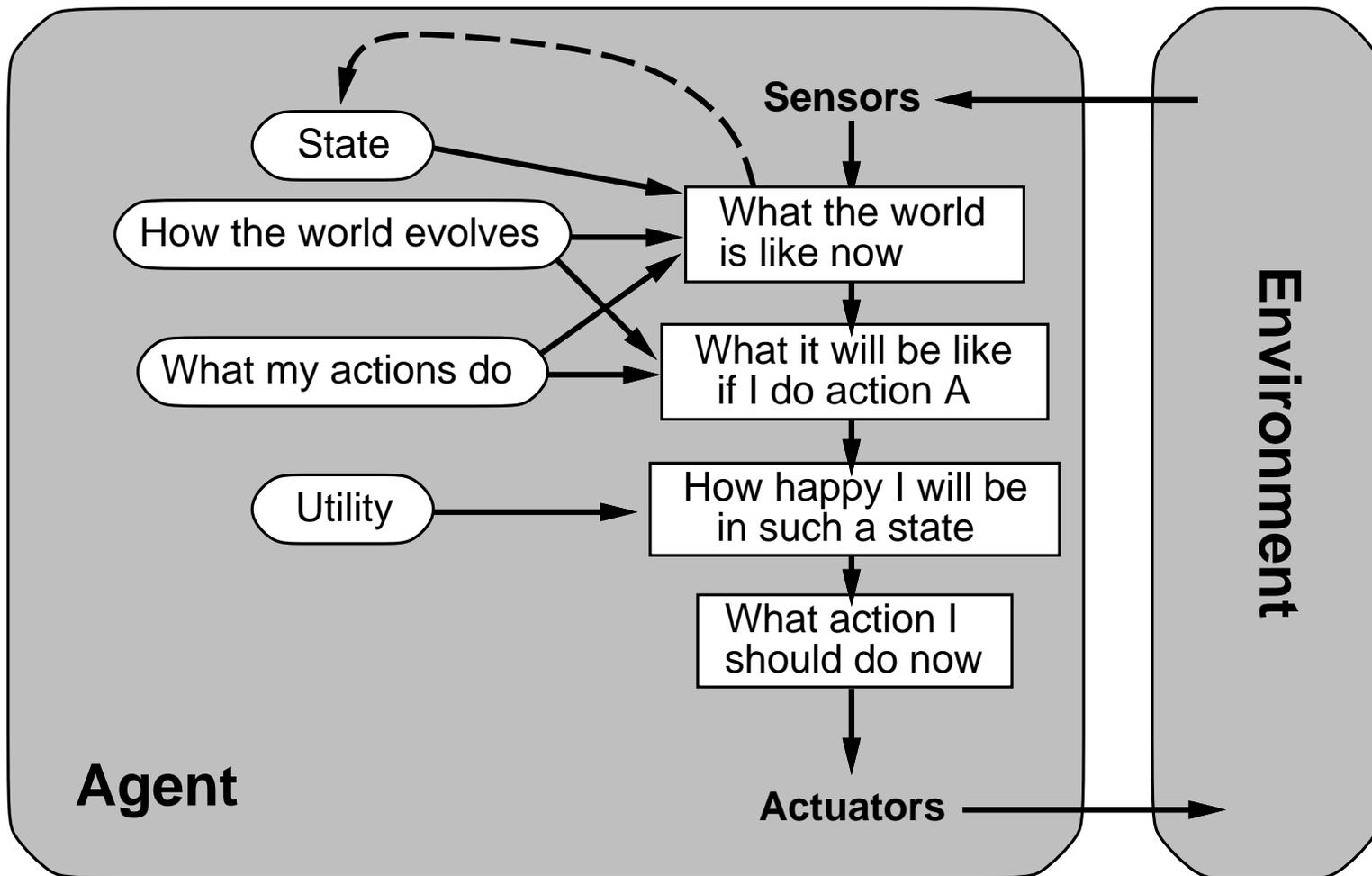
Example

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

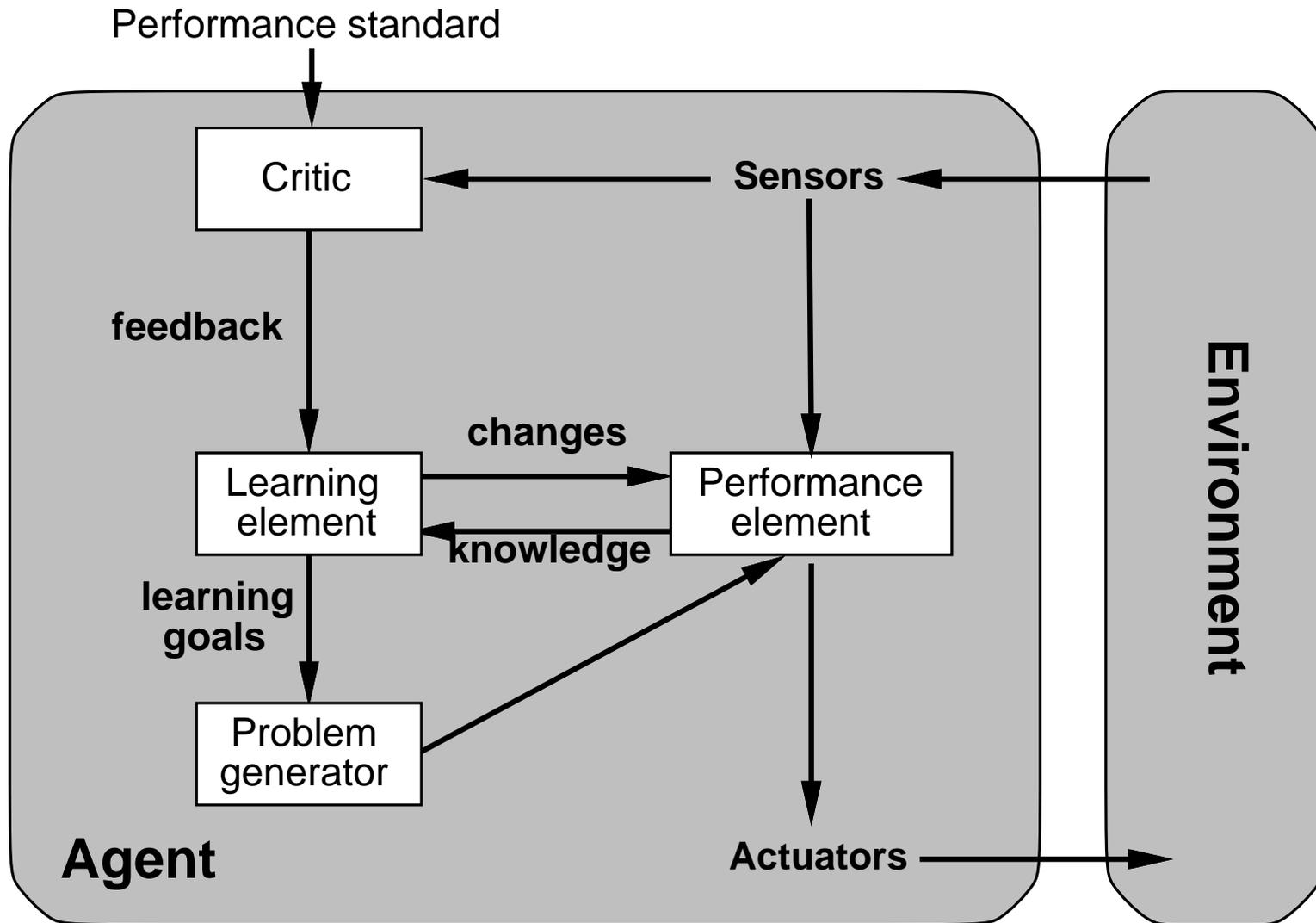
# Goal-based agents



# 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