



# Intelligent Agents

## Chapter 2



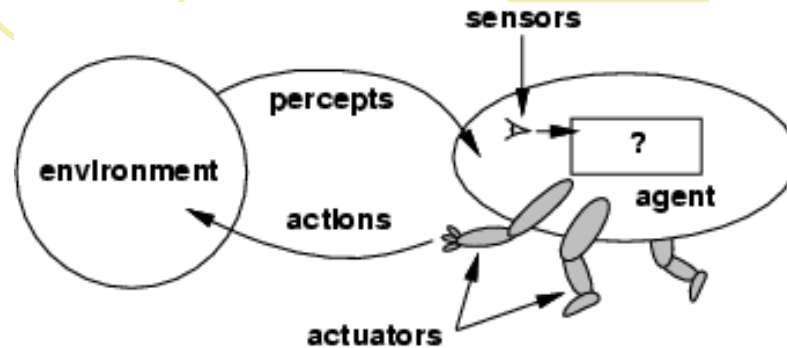
# Outline

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

# Agents

- An **agent** is anything that can be viewed as **perceiving** its **environment** through **sensors** and **acting** upon that environment through **actuators**
- Human agent: eyes, ears, and other organs for sensors; hands, legs, mouth, and other body parts for actuators
- Robotic agent: cameras and infrared range finders for sensors;
- various motors for actuators

# Agents and environments

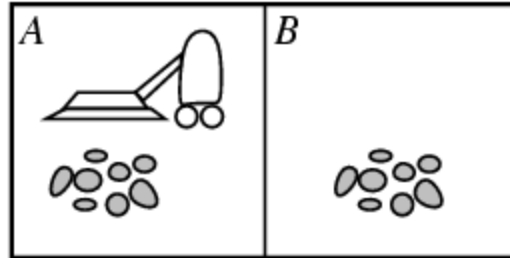


- The **agent function** maps from percept histories to actions:

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

- The **agent program** runs on the physical **architecture** to produce  $f$
- agent = architecture + program

# Vacuum-cleaner world



- Percepts: location and contents, e.g., [A,Dirty]
- Actions: *Left, Right, Suck, NoOp*

# A vacuum-cleaner agent

Percept sequence	Action
<i>[A, Clean]</i>	<i>Right</i>
<i>[A, Dirty]</i>	<i>Suck</i>
<i>[B, Clean]</i>	<i>Left</i>
<i>[B, Dirty]</i>	<i>Suck</i>
<i>[A, Clean], [A, Clean]</i>	<i>Right</i>
<i>[A, Clean], [A, Dirty]</i>	<i>Suck</i>
⋮	⋮

# Rational agents



- An agent should strive to "do the right thing", based on what it can perceive and the actions it can perform. The right action is the one that will cause the agent to be most successful
- Performance measure: An objective criterion for success of an agent's behavior
- E.g., performance measure of a vacuum-cleaner agent could be amount of dirt cleaned up, amount of time taken, amount of electricity consumed, amount of noise generated, etc.

# Rational agents



- **Rational Agent:** For each possible percept sequence, a rational agent should select an action that is expected to maximize its performance measure, given the evidence provided by the percept sequence and whatever built-in knowledge the agent has.



# Rational agents



- Rationality is distinct from omniscience (all-knowing with infinite knowledge)
- Agents can perform actions in order to modify future percepts so as to obtain useful information (information gathering, exploration)
- An agent is **autonomous** if its behavior is determined by its own experience (with ability to learn and adapt)



# PEAS

- PEAS: Performance measure, Environment, Actuators, Sensors
- Must first specify the setting for intelligent agent design
- Consider, e.g., the task of designing an automated taxi driver:
  - Performance measure
  - Environment
  - Actuators
  - Sensors

# PEAS



- Must first specify the setting for intelligent agent design
- Consider, e.g., the task of designing an automated taxi driver:
  - Performance measure: Safe, fast, legal, comfortable trip, maximize profits
  - Environment: Roads, other traffic, pedestrians, customers
  - Actuators: Steering wheel, accelerator, brake, signal, horn
  - Sensors: Cameras, sonar, speedometer, GPS, odometer, engine sensors, keyboard



PEAS

Agent: Medical diagnosis system

- Performance measure: Healthy patient, minimize costs, lawsuits
- Environment: Patient, hospital, staff
- Actuators: Screen display (questions, tests, diagnoses, treatments, referrals)
- Sensors: Keyboard (entry of symptoms, findings, patient's answers)



PEAS

Agent: Part-picking robot

- Performance measure: Percentage of parts in correct bins
- Environment: Conveyor belt with parts, bins
- Actuators: Jointed arm and hand
- Sensors: Camera, joint angle sensors



PEAS

Agent: Interactive English tutor

- Performance measure: Maximize student's score on test
- Environment: Set of students
- Actuators: Screen display (exercises, suggestions, corrections)
- Sensors: Keyboard

# Environment types



- **Fully observable** (vs. partially observable): An agent's sensors give it access to the complete state of the environment at each point in time.
- **Deterministic** (vs. stochastic): The next state of the environment is completely determined by the current state and the action executed by the agent. (If the environment is deterministic except for the actions of other agents, then the environment is **strategic**)
- **Episodic** (vs. sequential): The agent's experience is divided into atomic "episodes" (each episode consists of the agent perceiving and then performing a single action), and the choice of action in each episode depends only on the episode itself.

# Environment types



- **Static** (vs. dynamic): The environment is unchanged while an agent is deliberating. (The environment is **semidynamic** if the environment itself does not change with the passage of time but the agent's performance score does)
- **Discrete** (vs. continuous): A limited number of distinct, clearly defined percepts and actions.
- **Single agent** (vs. multiagent): An agent operating by itself in an environment.



# Environment types

	Chess with a clock	Chess without a clock	Taxi driving
Fully observable	Yes	Yes	No
Deterministic	Strategic	Strategic	No
Episodic	No	No	No
Static	Semi	Yes	No
Discrete	Yes	Yes	No
Single agent	No	No	No

- The environment type largely determines the agent design
- The real world is (of course) partially observable, stochastic, sequential, dynamic, continuous, multi-agent

# Agent functions and programs

- An agent is completely specified by the agent function mapping percept sequences to actions
- One agent function (or a small equivalence class) is rational
- Aim: find a way to implement the rational agent function concisely

# Table-lookup agent

```
function TABLE-DRIVEN-AGENT(percept) returns action  
  static: percepts, a sequence, initially empty  
           table, a table of actions, indexed by percept sequences, fully specified  
  
  append percept to the end of percepts  
  action ← LOOKUP(percepts, table)  
  return action
```

## ● Drawbacks:

- Huge table
- Take a long time to build the table
- No autonomy
- Even with learning, need a long time to learn the table entries

# Agent program for a vacuum-cleaner agent

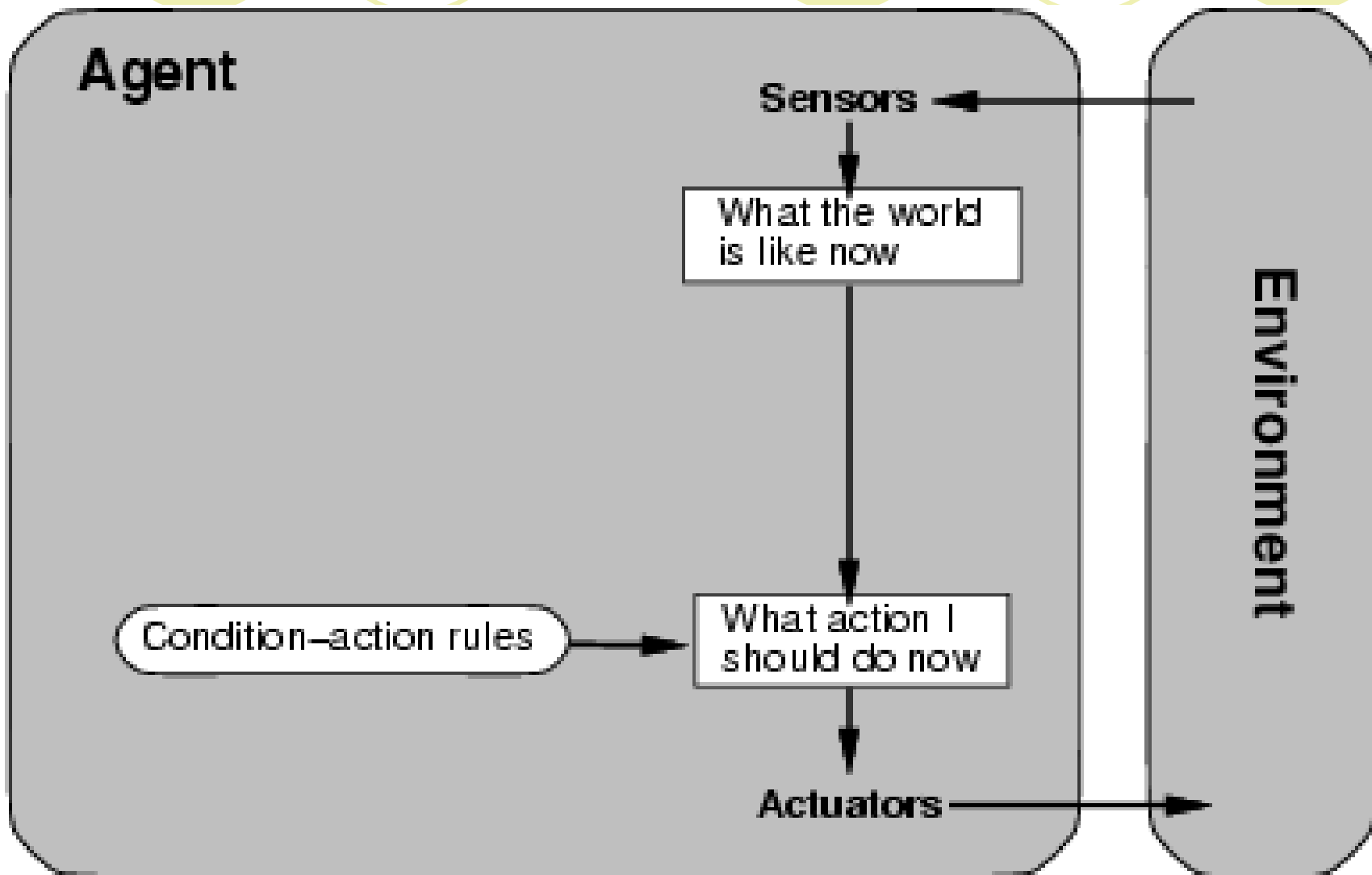
```
function REFLEX-VACUUM-AGENT( [location, status]) returns an action  
  if status = Dirty then return Suck  
  else if location = A then return Right  
  else if location = B then return Left
```

# Agent types



- Four basic types in order of increasing generality:
- Simple reflex agents
- Model-based reflex agents
- Goal-based agents
- Utility-based agents

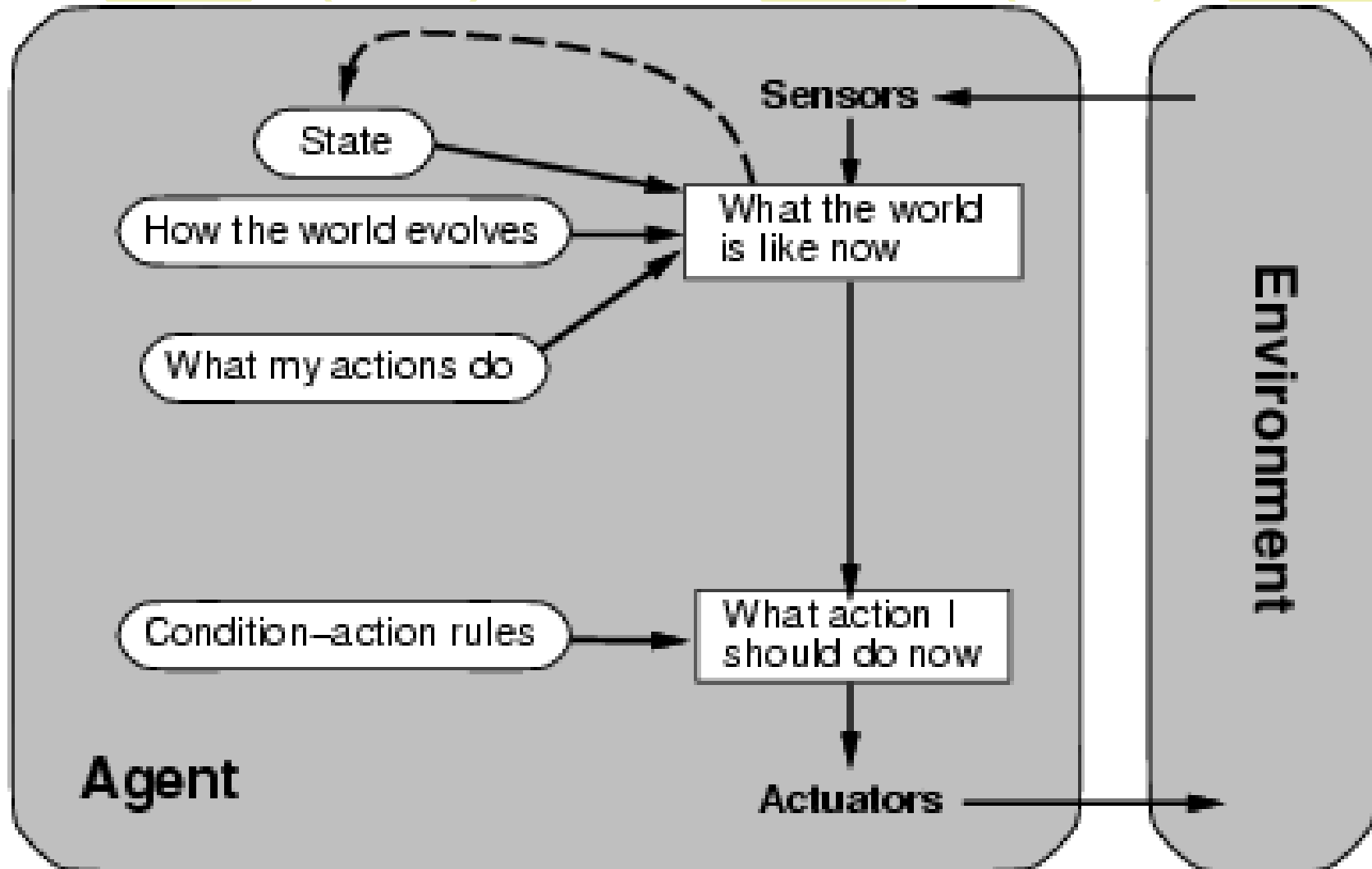
# Simple reflex agents



# Simple reflex agents

```
function SIMPLE-REFLEX-AGENT(percept) returns action  
  static: rules, a set of condition-action rules  
  
  state ← INTERPRET-INPUT(percept)  
  rule ← RULE-MATCH(state, rules)  
  action ← RULE-ACTION[rule]  
  return action
```

# Model-based reflex agents

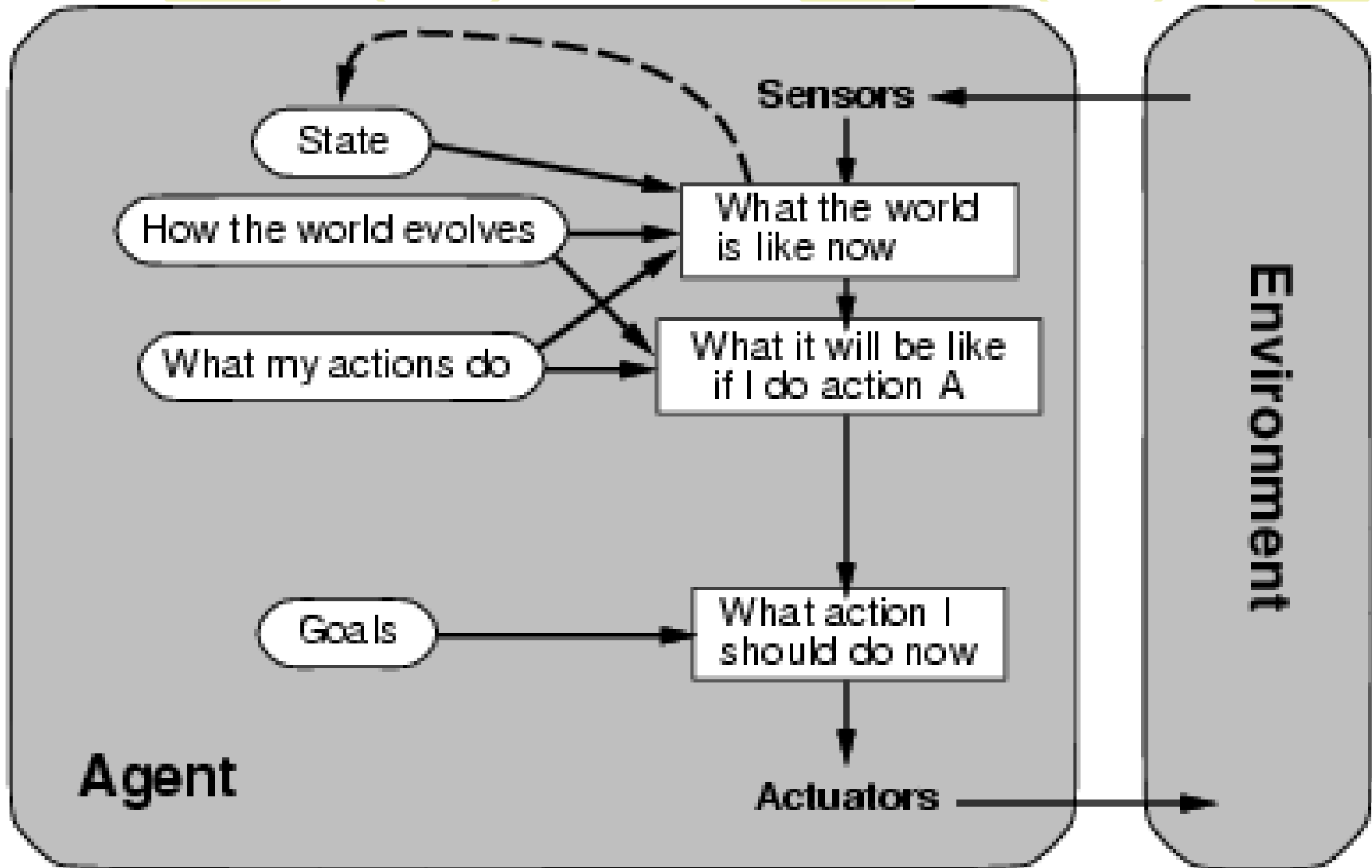




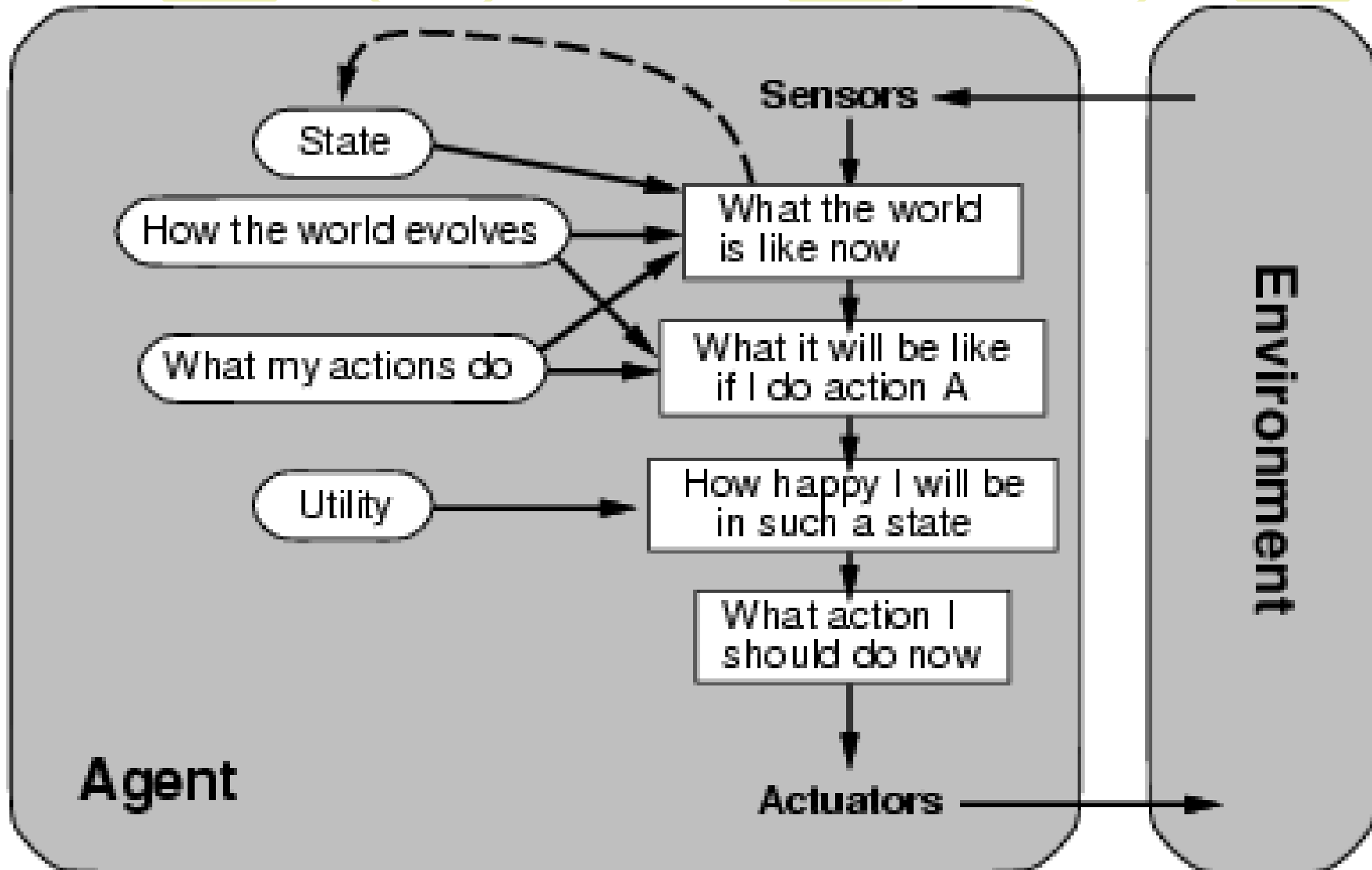
# Model-based reflex agents

```
function REFLEX-AGENT-WITH-STATE(percept) returns action  
  static: state, a description of the current world state  
           rules, a set of condition-action rules  
           action, the most recent action, initially none  
  
  state ← UPDATE-STATE(state, action, percept)  
  rule ← RULE-MATCH(state, rules)  
  action ← RULE-ACTION[rule]  
  return action
```

# Goal-based agents



# Utility-based agents



# Learning agents

