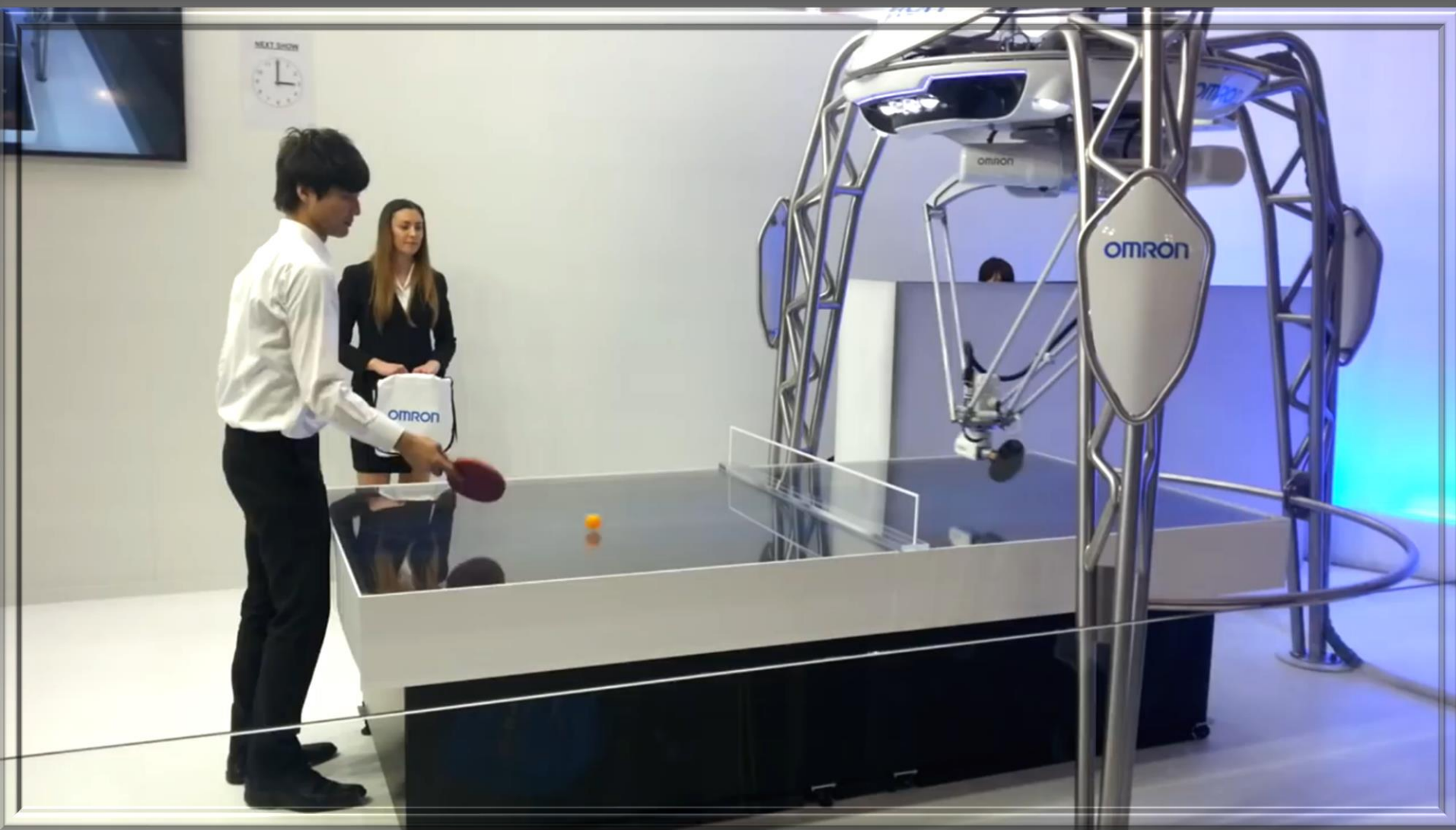


Lecture #3

Outline

- Agents and environments
- Rationality
- **PEAS** (**P**erformance **M**easure, **E**nvironment, **A**ctuators, **S**ensors)
- Environment types
- Agent types

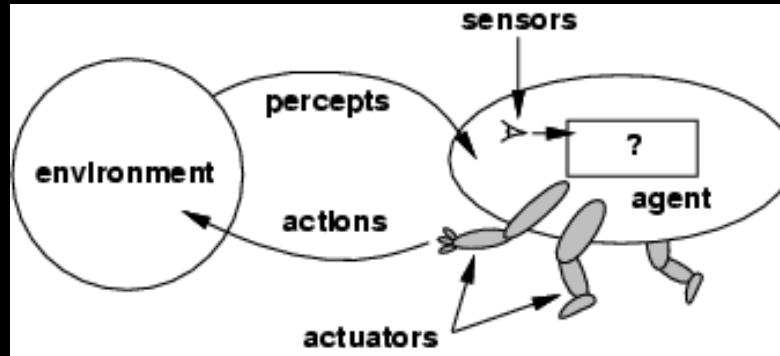
Question?



Driving in downtown Cairo is too unpredictable

- 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:

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

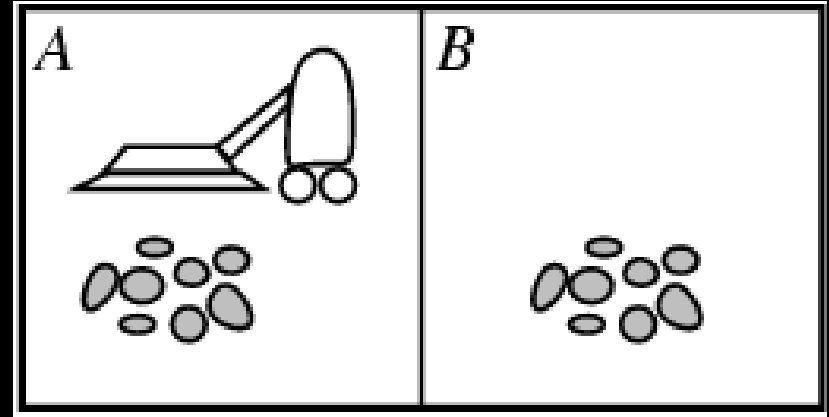
The **agent program** runs on the physical **architecture** to produce f

agent = architecture + program

Vacuum-cleaner world

- **Percepts:**

Location and status,
e.g., [A,Dirty]



- **Actions:**

Left, Right, Suck, NoOp

function 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*

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**.

•

An agent acts intelligently when

- what it does is appropriate for its functions and its goals,
 - it is flexible to changing environments and changing goals,
 - it learns from experience, and
 - it makes appropriate choices given its perceptual and computational limitations.
 - An agent typically cannot observe the state of the world directly; it has only a finite memory and it does not have unlimited time to act.
- A **computational agent** is an agent whose **decisions** about its **actions** can be **explained** in terms of computation.
- the decision can be broken down into primitive operation that can be implemented in a physical device.

Knowledge is the information about a domain that can be used to solve problems in that domain.

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

An Autonomous Delivery Robot



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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

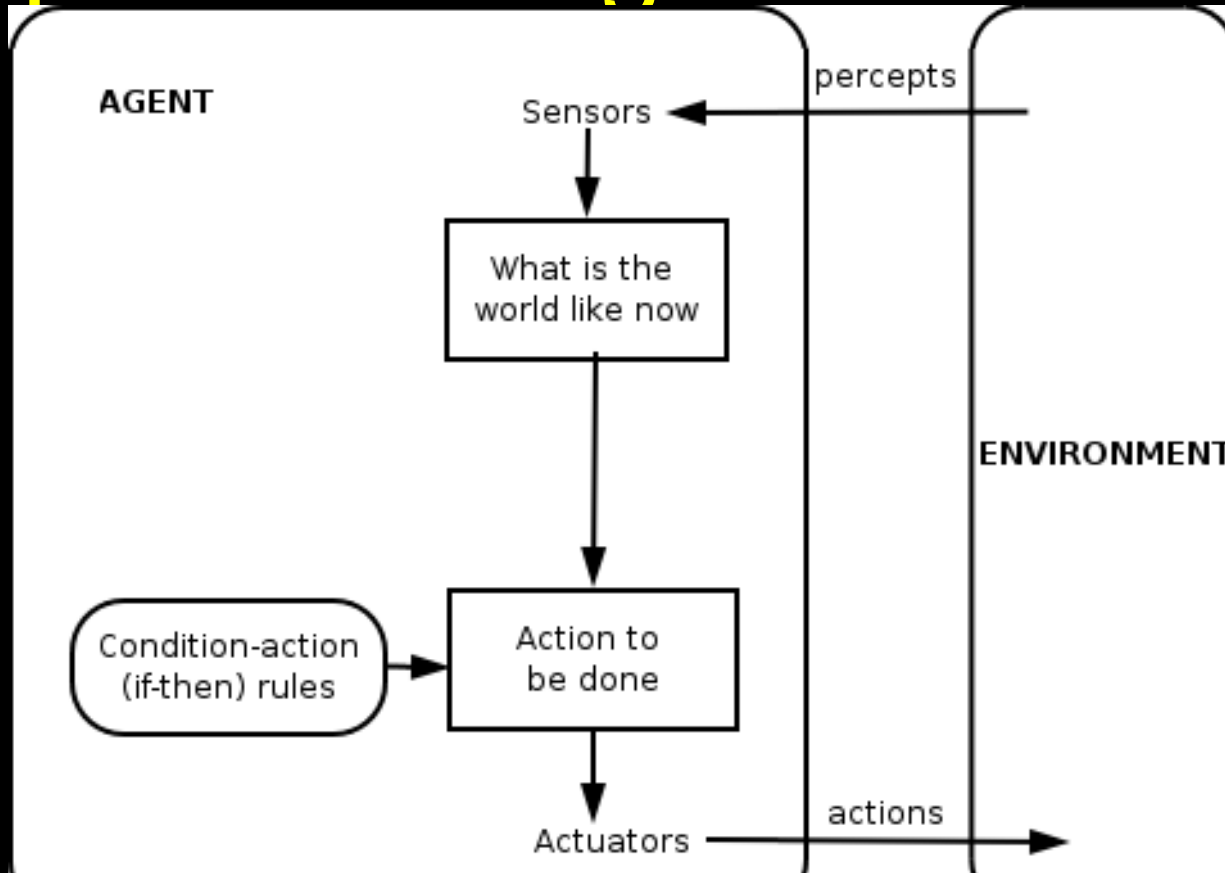
- ✓ **Static** (vs. dynamic): The environment is unchanged while an agent is deliberating. (The environment is **semidynamic** if the environment itself does not change with the passage of time but the agent's performance score does)
- ✓
- ✓ **Discrete** (vs. continuous): A limited number of distinct,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



Agent types

- Simple reflex agents
- Model-based reflex agents
- Goal-based agents
- Utility-based agents

Simple reflex agents



A simple
current

function SIMPLE-REFLEX-AGENT(*percept*) **returns** an action

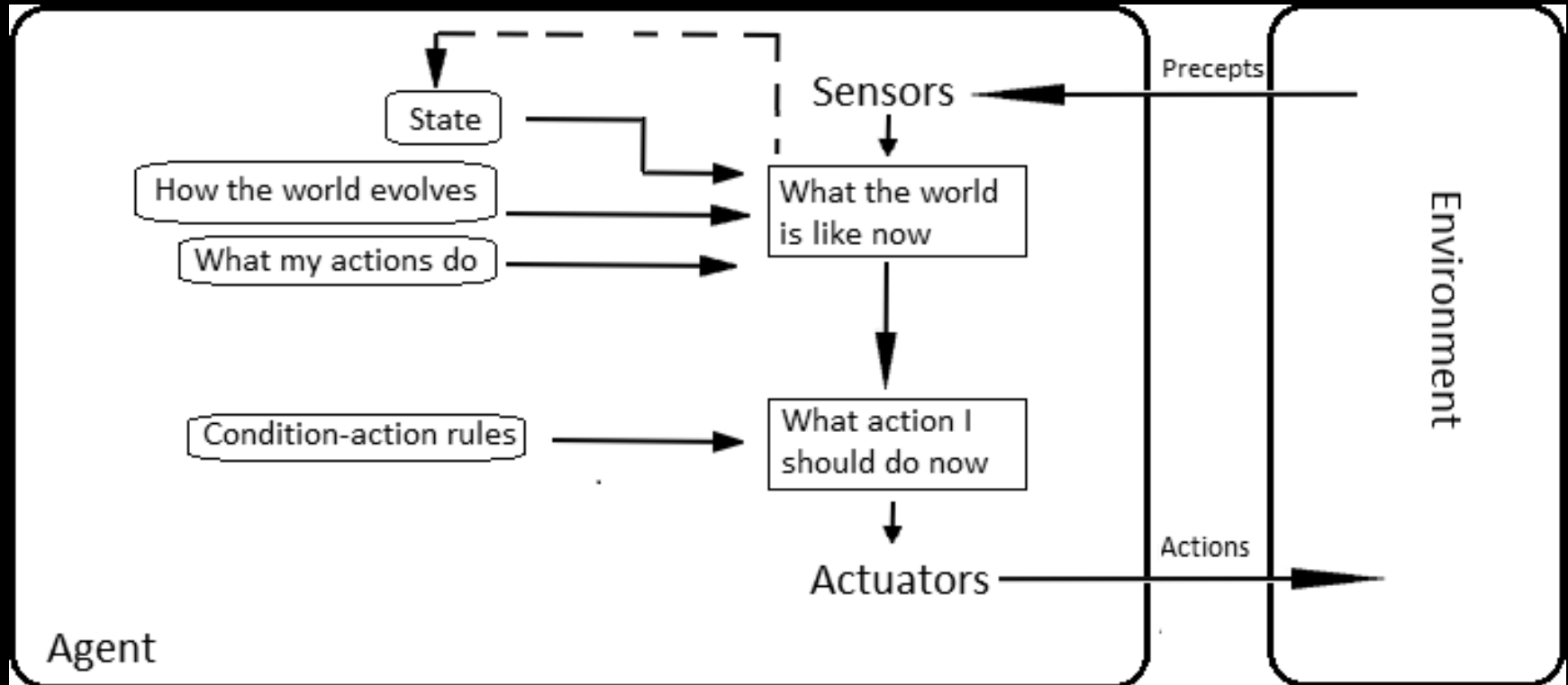
persistent: *rules*, a set of condition–action rules

state \leftarrow INTERPRET-INPUT(*percept*) *rule* \leftarrow RULE-MATCH(*state*,
rules) *action* \leftarrow *rule*.ACTION

return *action*

has the

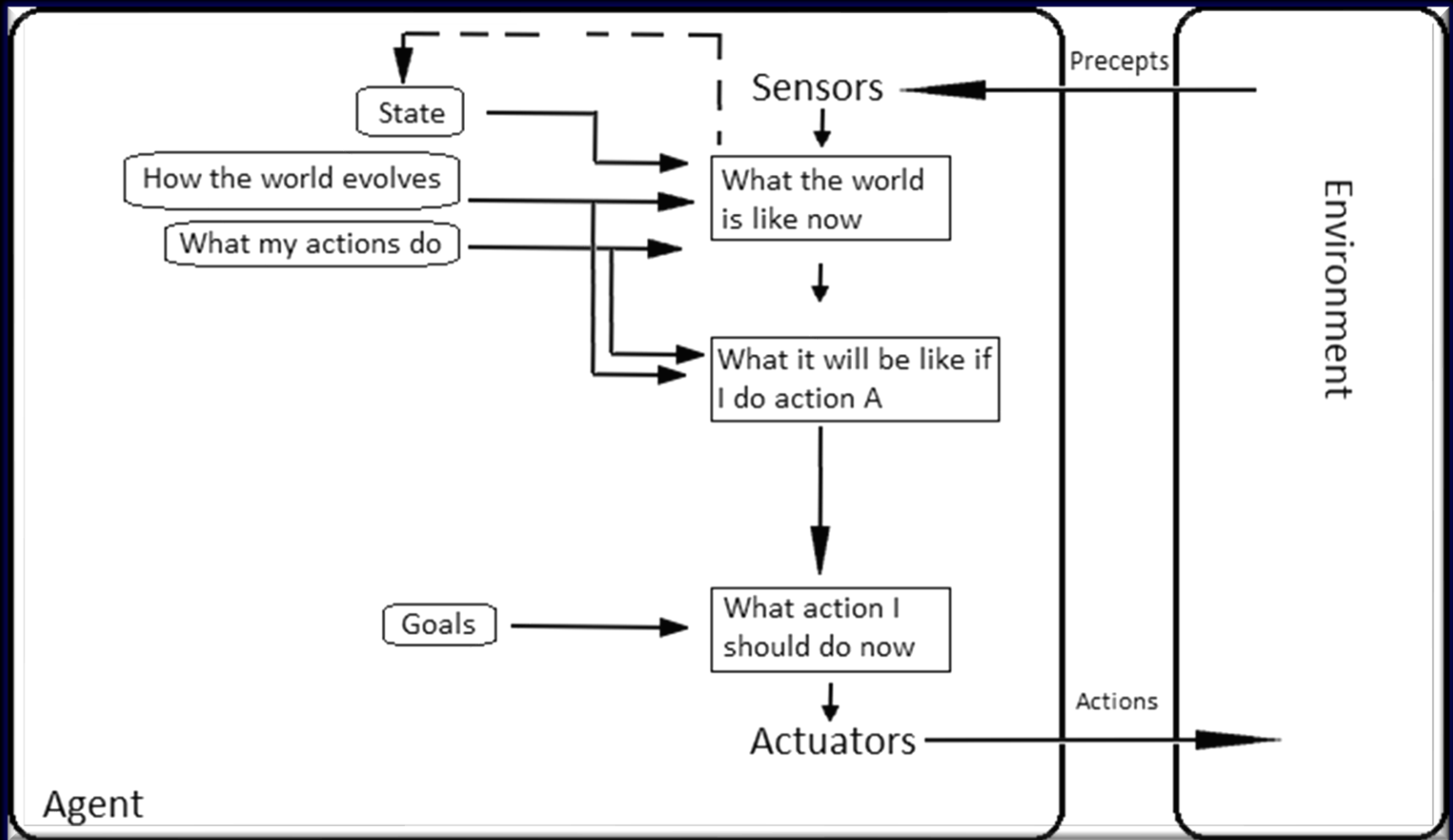
Model-based reflex agents



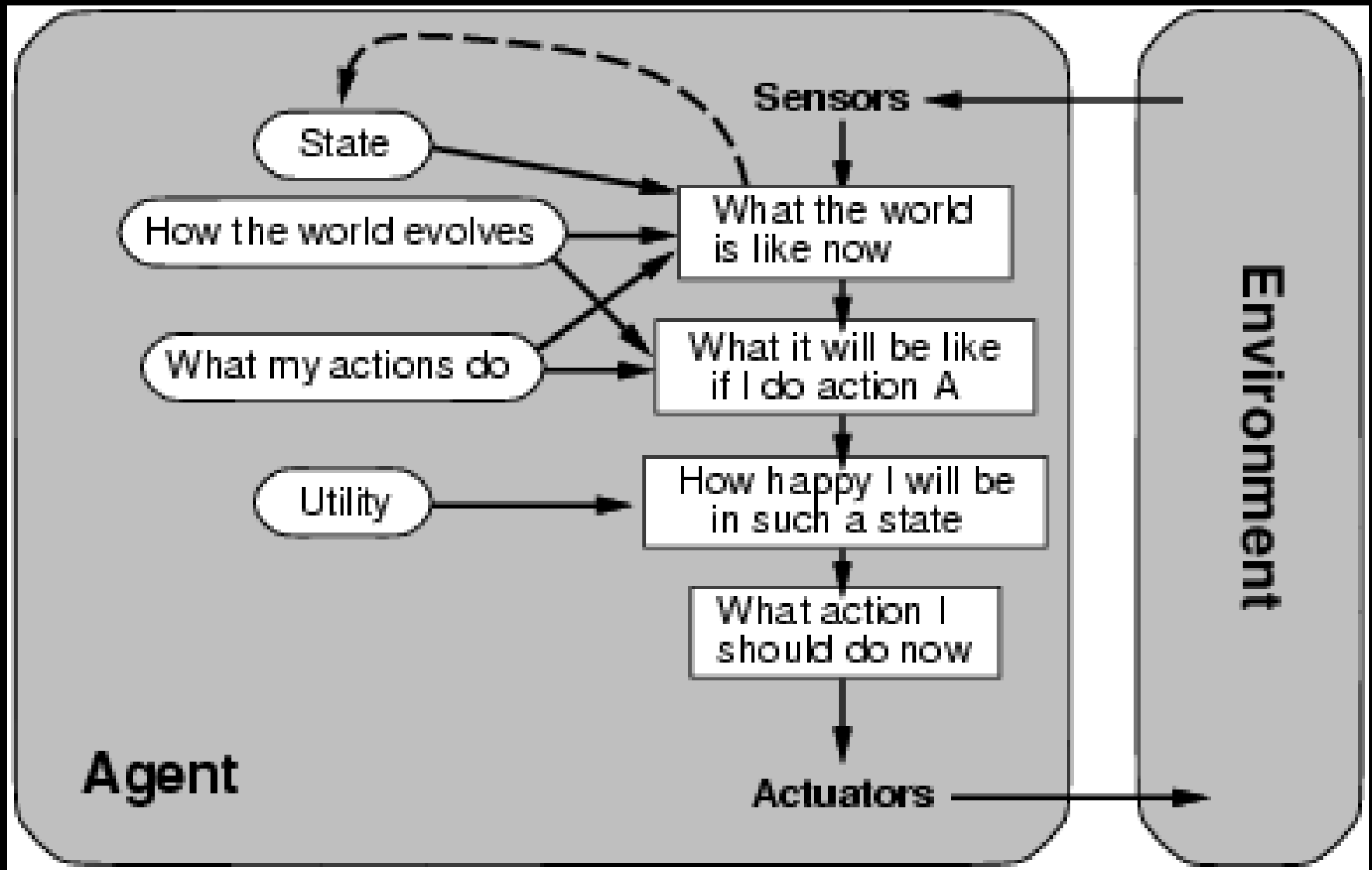
the agent should maintain some sort of **internal** state that depends on the **percept history** and thereby **reflects** at least some of the unobserved aspects of the current state.

For the **braking** problem, the **internal state** is not too extensive—just the previous frame from the camera, allowing the **agent** to **detect** when **two red lights** at the edge of the vehicle go on or off simultaneously. For other driving tasks such as changing lanes, the agent needs to keep track of where the other cars are if it can't see them all at once. And for any driving to be possible at all, the agent needs to keep track of where its keys are.

Goal-based agents



Utility-based agents



Learning agents

