



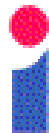
Intelligent Agents and their Environments

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Structure of Intelligent Agents

An agent:

- Perceives its *environment*,
- Through its *sensors*,
- Then achieves its *goals*
- By acting on its environment via *actuators*.

Examples of Agents 1

- **Agent**: mail sorting robot
- **Environment**: conveyor belt of letters
- **Goals**: route letter into correct bin
- **Percepts**: array of pixel intensities
- **Actions**: route letter into bin

Examples of Agents 2

- **Agent:** intelligent house
- **Environment:**
 - occupants enter and leave house,
 - occupants enter and leave rooms;
 - daily variation in outside light and temperature
- **Goals:** occupants warm, room lights are on when room is occupied, house energy efficient
- **Percepts:** signals from temperature sensor, movement sensor, clock, sound sensor
- **Actions:** room heaters on/off, lights on/off

Examples of Agents 3

- **Agent**: automatic car.
- **Environment**: streets, other vehicles, pedestrians, traffic signals/lights/signs.
- **Goals**: safe, fast, legal trip.
- **Percepts**: camera, GPS signals, speedometer, sonar.
- **Actions**: steer, accelerate, brake.

Side info: http://en.wikipedia.org/wiki/2005_DARPA_Grand_Challenge

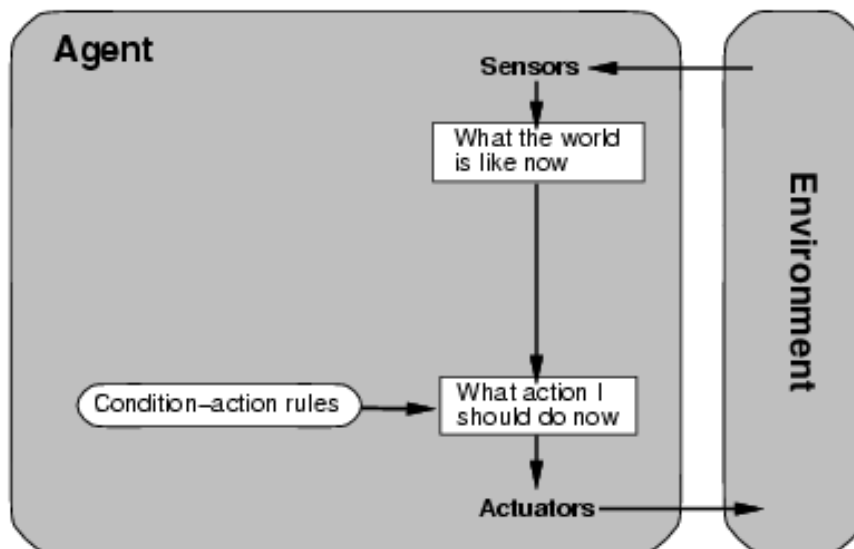
Simple Reflex Agents

- Action depends only on immediate percepts.
- Implement by *condition-action rules*.

Example:

- **Agent**: Mail sorting robot
- **Environment**: Conveyor belt of letters
- **Rule**: e.g. *city=Edin* → *put Scotland bag*

Simple Reflex Agents



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

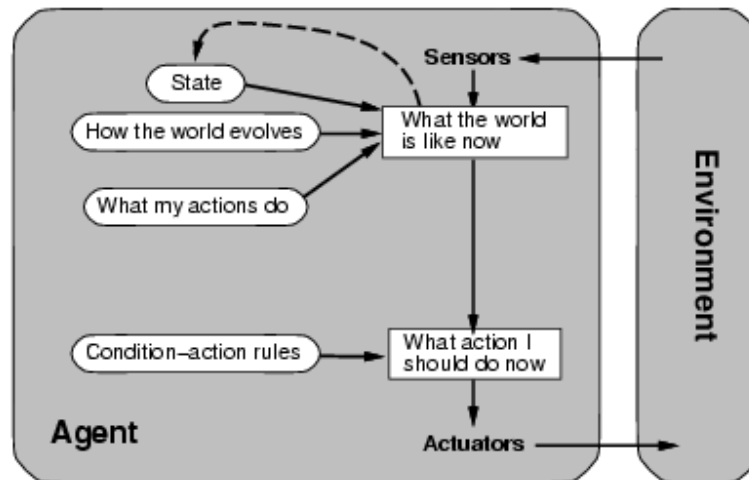
Model-Based Reflex Agents

- Action may depend on history or unperceived aspects of the world.
- Need to maintain *internal world model*.

Example:

- **Agent**: robot vacuum cleaner
- **Environment**: dirty room, furniture.
- **Model**: map of room, which areas already cleaned.
- Sensor/model tradeoff.

Model-Based Reflex Agents



function REFLEX-AGENT-WITH-STATE(*percept*)

returns *action*

persistent: *state*, description of current world state

model, description of how the next state depends on
current state and action

rules, a set of condition-action rules

action, the most recent action, initially none

state ← UPDATE-STATE(*state*, *action*, *percept*, *model*)

rule ← RULE-MATCH(*state*, *rules*)

action ← rule.ACTION

return *action*

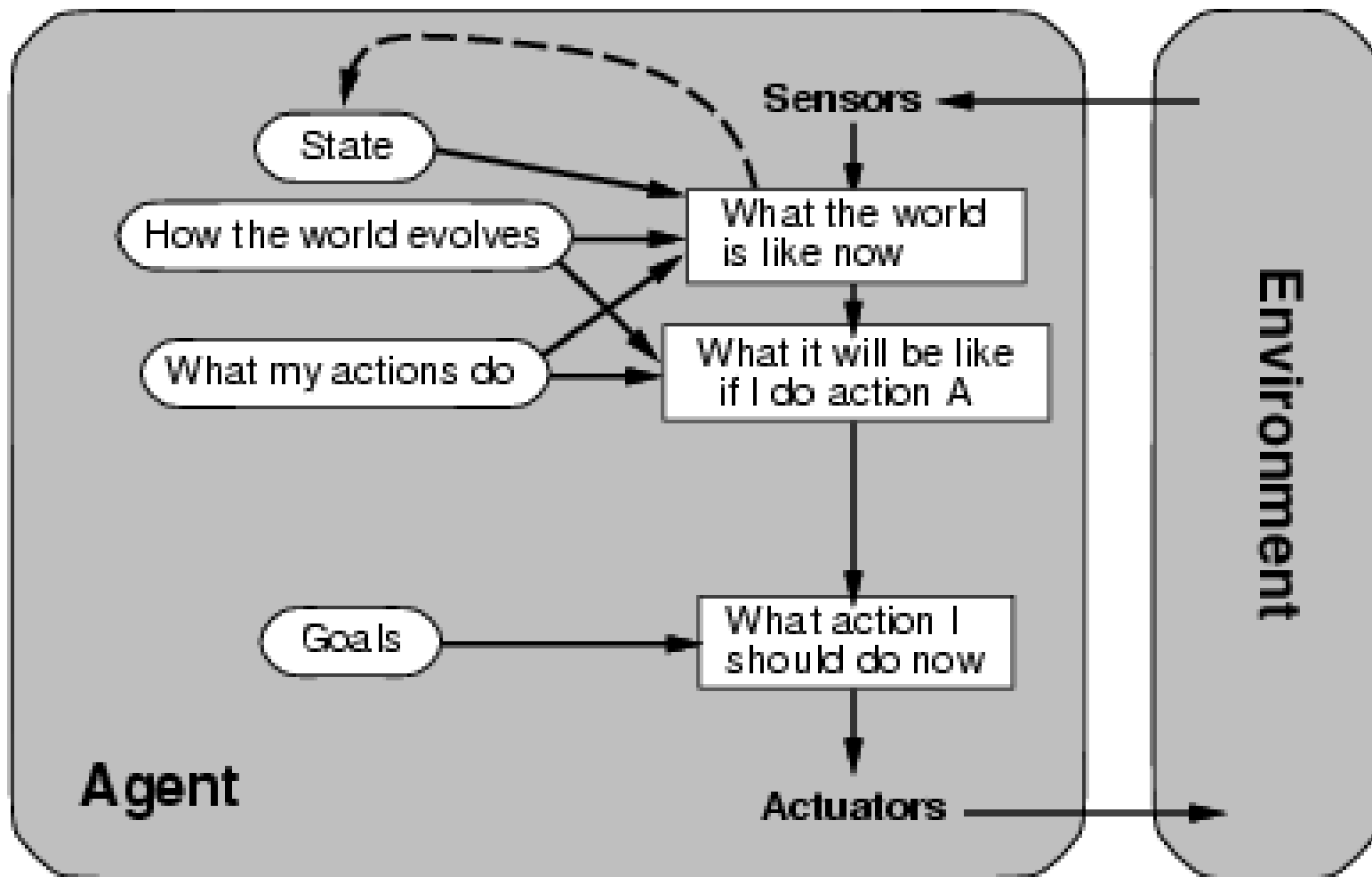
Goal-Based Agents

- Agents so far have fixed, implicit goals.
- We want agents with variable goals.
- Forming plans to achieve goals is later topic.

Example:

- **Agent**: robot maid
- **Environment**: house & people.
- **Goals**: clean clothes, tidy room, table laid, etc

Goal-Based Agents



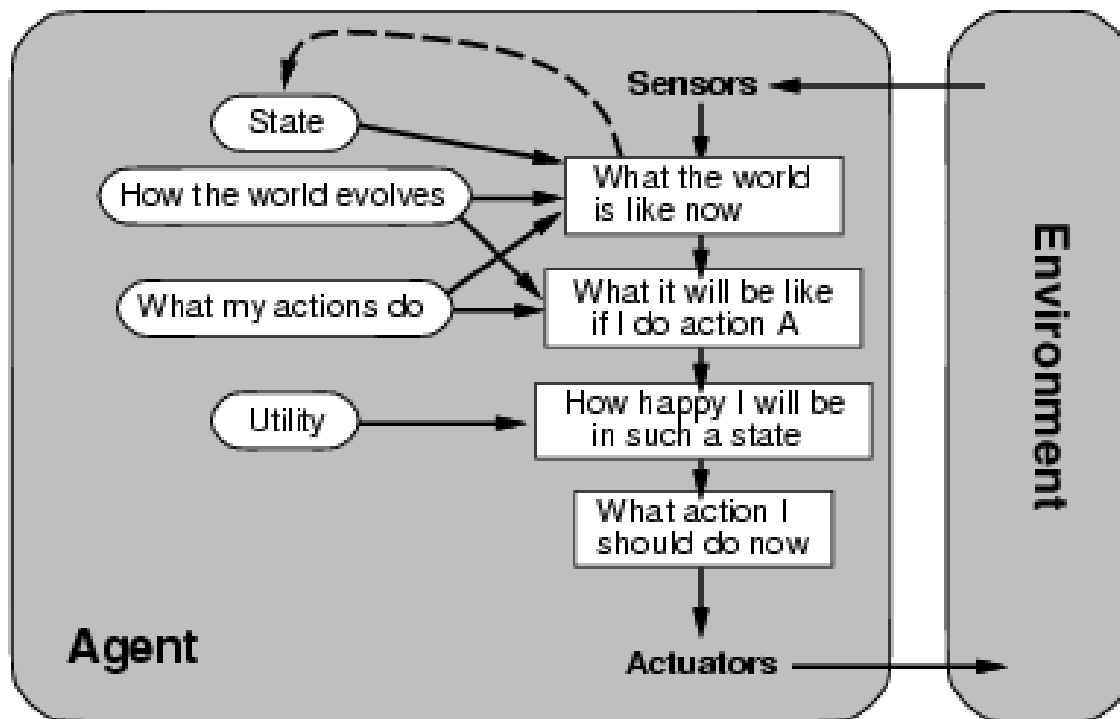
Utility-Based Agents

- Agents so far have had a single goal.
- Agents may have to juggle conflicting goals.
- Need to optimise utility over a range of goals.
- **Utility**: measure of *goodness* (a real number).
- Combine with probability of success to get *expected utility*.

Example:

- **Agent**: automatic car.
- **Environment**: roads, vehicles, signs, etc.
- **Goals**: stay safe, reach destination, be quick, obey law, save fuel, etc.

Utility-Based Agents



We will not be covering utility-based agents, but this topic is discussed in Russell & Norvig, Chapters 16 and 17

Learning Agents

How do agents improve their performance in the light of experience?

- Generate problems which will test performance.
- Perform activities according to rules, goals, model, utilities, etc.
- Monitor performance and identify non-optimal activity.
- Identify and implement improvements.

We will not be covering learning agents, but this topic is discussed in Russell & Norvig, Chapters 18-21.

Mid Lecture Exercise

Consider a chess playing program.
What sort of agent would it need to be?

Solution

- **Simple-reflex agent:** but some actions require some memory (e.g. castling in chess - <http://en.wikipedia.org/wiki/Castling>).
- **Model-based reflex agent:** but needs to reason about future.
- **Goal-based agent:** but only has one goal.
- **Utility-based agent:** might consider multiple goals with limited lookahead.

Types of Environment 1

- **Fully Observable vs. Partially Observable:**
Observable: agent's sensors describe environment fully.
Playing chess with a blindfold.
- **Deterministic vs. Stochastic:**
Deterministic: next state fully determined by current state and agent's actions.
Chess playing in a strong wind.

An environment may appear stochastic if it is only partially observable.

Types of Environment 2

- **Episodic vs. Sequential:**

Episodic: next episode does not depend on previous actions.

Mail-sorting robot *vs* crossword puzzle.

- **Static vs. Dynamic:**

Static: environment unchanged while agent deliberates.

Robot car *vs* chess.

Crossword puzzle *vs* tetris.

Types of Environment 3

- **Discrete vs. Continuous:**
Discrete: percepts, actions and episodes are discrete.
Chess *vs* robot car.
- **Single Agent vs. Multi-Agent:**
How many objects must be modelled as agents.
Crossword *vs* poker.

Element of choice over which objects are considered agents.

Types of Environment 4

- An agent might have any combination of these properties:
 - from “benign” (i.e., fully observable, deterministic, episodic, static, discrete and single agent)
 - to “chaotic” (i.e., partially observable, stochastic, sequential, dynamic, continuous and multi-agent).
- What are the properties of the environment that would be experienced by
 - a mail-sorting robot?
 - an intelligent house?
 - a car-driving robot?

Summary

- Simple reflex agents
- Model-based reflex agents
- Goal-based agents
- Utility-based agents
- Learning agents
- Properties of environments