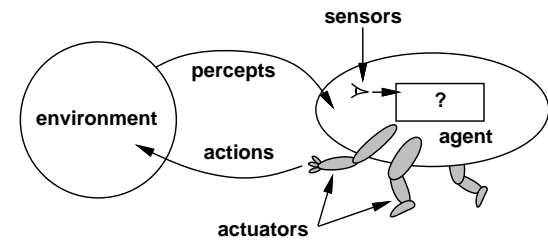


INTELLIGENT AGENTS

CHAPTER 2

Agents and environments



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

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

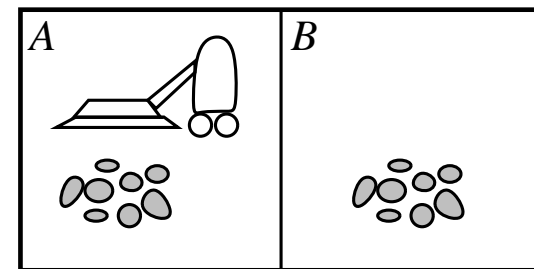
$$f : \mathcal{P}^* \rightarrow \mathcal{A}$$

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

Outline

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

Vacuum-cleaner world



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

Actions: *Left, Right, Suck, NoOp*

A vacuum-cleaner agent

Percept sequence	Action
[A, Clean]	Right
[A, Dirty]	Suck
[B, Clean]	Left
[B, Dirty]	Suck
[A, Clean], [A, Clean]	Right
[A, Clean], [A, Dirty]	Suck
⋮	⋮

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/correct” function?

Can it be implemented in a small agent program?

Chapter 2 5

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??

Chapter 2 7

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

Chapter 2 6

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, ...

Chapter 2 8

Internet shopping agent

Performance measure??

Environment??

Actuators??

Sensors??

Chapter 2 9

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)

Chapter 2 10

Environment types

Fully observable vs. partially observable

– Can the agent observe/know everything in a state?

Deterministic vs. stochastic

– Does the current state plus action fully determines the next state?

Episodic vs. sequential

– Does the action affect the future action(s)?

– Going to class does not affect doing homework in the future.

– How you make a move in a chess game affects your moves later.

Static vs. dynamic

– Can the environment change while the agent is thinking?

Discrete vs. continuous

– Finitely distinct or infinitely continuous?

Single agent vs. multi-agent – Does the agent deal with other agents?

Chapter 2 11

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>				

Chapter 2 12

Environment types

	Solitaire	Backgammon	Internet shopping	Taxi
<u>Observable??</u>	Yes	Yes	No	No
<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>				
<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
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<u>Discrete??</u>				
<u>Single-agent??</u>				

Environment types

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<u>Static??</u>	Yes	Semi	Semi	No
<u>Discrete??</u>				
<u>Single-agent??</u>				

Environment types

	Solitaire	Backgammon	Internet shopping	Taxi
Observable??	Yes	Yes	No	No
Deterministic??	Yes	No	Partly	No
Episodic??	No	No	No	No
Static??	Yes	Semi	Semi	No
Discrete??	Yes	Yes	Yes	No
Single-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

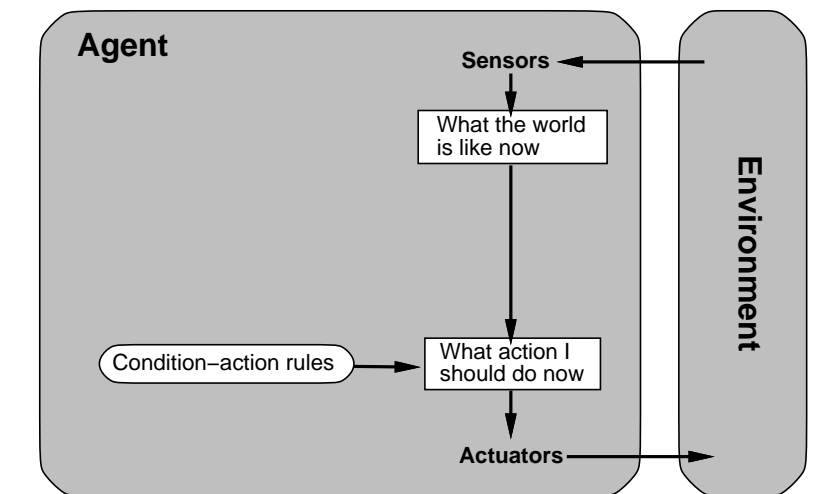
Environment types

	Solitaire	Backgammon	Internet shopping	Taxi
Observable??	Yes	Yes	No	No
Deterministic??	Yes	No	Partly	No
Episodic??	No	No	No	No
Static??	Yes	Semi	Semi	No
Discrete??	Yes	Yes	Yes	No
Single-agent??	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

Simple reflex agents



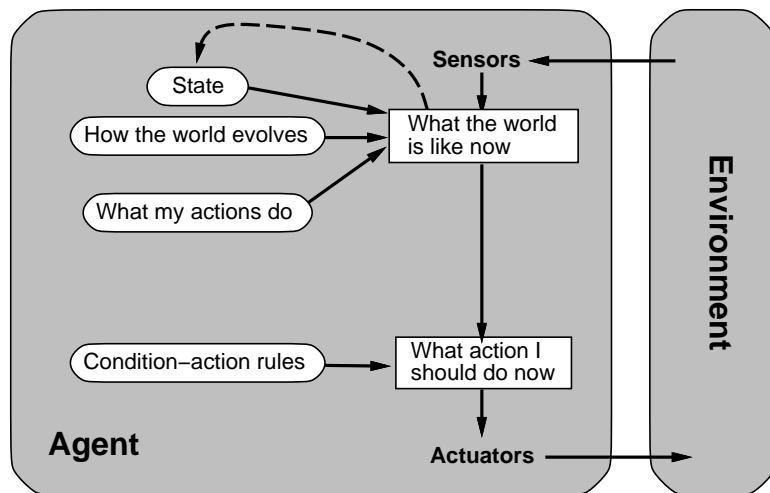
Example

```
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
```

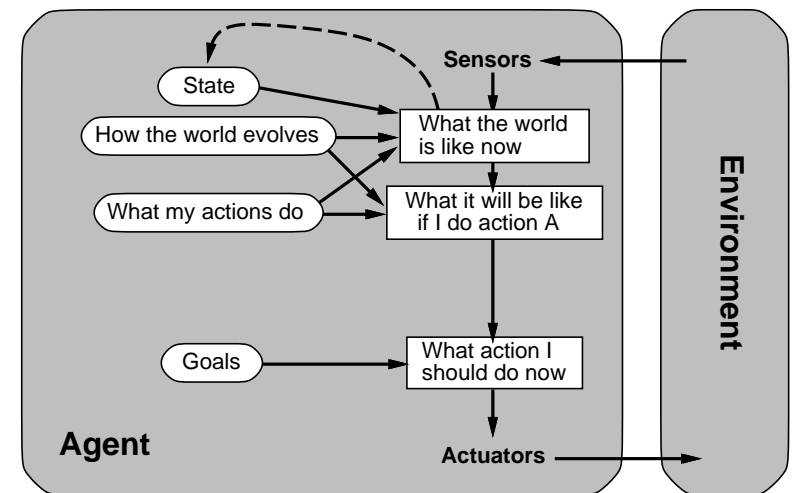
Example

```
function REFLEX-VACUUM-AGENT([location, status]) returns an action
  state ← UPDATE-STATE(state, location, status)
  if state = ... AND status = Dirty then ...
```

Reflex agents with state

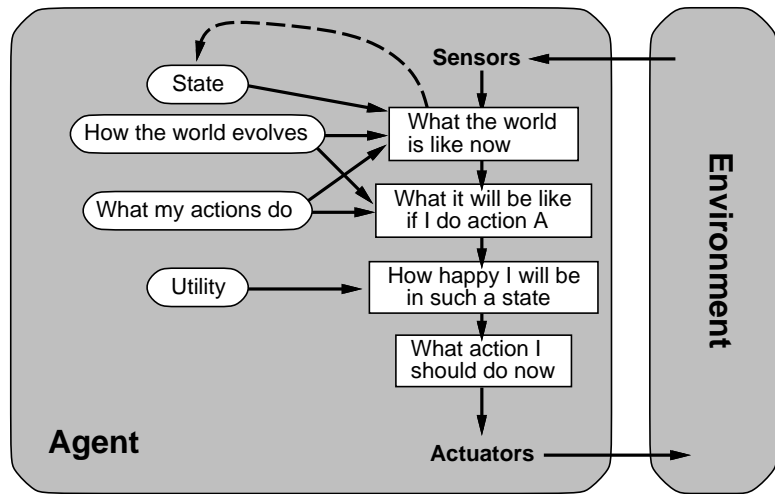


Goal-based agents



Why add an internal model of how the environment evolves?

Utility-based 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

Learning agents

