

Artificial Intelligence

2. Intelligent Agents

Let's Get a Little Orientation Here

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Agenda

- 1 Introduction
- 2 What is an Agent?
- 3 What is a Rational Agent?
- 4 Different Classes of Agents
- 5 Different Classes of Environments
- 6 Conclusion

Introduction

→ A central aspect of intelligence (and one possible way to define it) is the ability to act successfully in the world (cf. **Chapter 1**).

This chapter provides a broad categorization of and terminology for what we mean by “acting successfully”.

- Useful for a research field to establish a common language.
- Useful for you to get a rough overview.

→ We'll think a bit about what it means to “behave rationally”, what are possible architectures for achieving such behavior with a computer, and what are relevant properties of the world we need to act in.

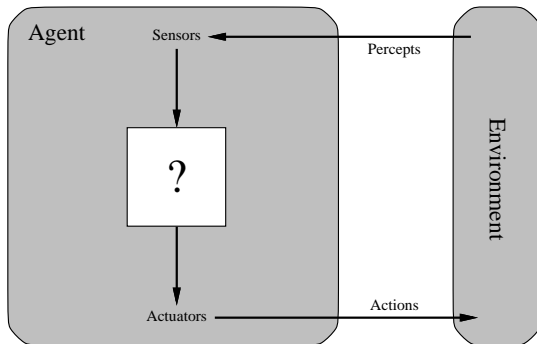
Disclaimer

→ The concepts and explanations in this chapter are very broad and rather superficial. In the remainder of the course, we will consider in detail particular problems relating to decision-making and specify algorithms for solving these.

What is an “Agent” in AI?

Agents:

- Perceive the environment through **sensors** (→ **percepts**).
- Act upon the environment through **actuators** (→ **actions**).



→ Examples?

Rational Agents ...

... do “the right thing”!

→ Meaning of “do the right thing”: Rational agents select their actions so as to maximize a performance measure.

→ What’s the performance measure of an autonomous vacuum cleaner?

Actually, Rational Agents ...

... ATTEMPT to do “the right thing”!

→ The hypothetical best case (“the right thing”) is often unattainable.

→ The agent might not be equipped with the right action to perform.
(Vacuum cleaner cannot step down the stairs.)

→ The agent might not be able to perceive all relevant information.
(Is there dirt under this bed? Is this a hamster or a shoe?)

Rationality vs. Omniscience:

- An **omniscient agent** knows everything about the environment and knows the actual effects of its actions.
- A **rational agent** just makes the best of what it has at its disposal, *maximizing expected performance given its percepts, knowledge and available actions.*

→ Example?

So, What *Is* a Rational Agent?

Mapping input to the best possible output:

Performance measure \times Percepts \times Knowledge \rightarrow Action

- An **agent** has a performance measure M and a set A of possible actions. Given a percept sequence P , as well as knowledge K about the world, it selects an action $a \in A$.
- The action a is **optimal** if it maximizes the expected value of M , given the evidence provided by P and K . The agent is **rational** if it always chooses an optimal a .

\rightarrow If the vacuum cleaner bumps into the hamster, then this can be rational in case

\rightarrow Note: If **observation actions** are required, they are elements of A , i.e., the agent must *perceive actively*. **Example:**

A Rational Agent is an Optimal Action Choice Function?

We also need to realize the agent through:

- an **agent program**, executed on
- an **architecture** which also provides an interface to the environment (percepts, actions).

→ Agent = Architecture + Program

Practical limitations:

- Our definition captures limitations on percepts and knowledge.
- It does not capture computational limitations (often, determining an optimal choice would take too much time/memory).
 - In practice, we often merely *approximate* the rational decision.

Examples of Agents: PEAS Descriptions

Questionnaire

Question!

Which are agents?

(A): James Bond.

(B): Your dog.

(C): Vacuum cleaner.

(D): Thermometer.

Question!

Who is rational?

(A): James Bond, crossing the street without looking.

(B): Your dog, crossing the street without looking.

(C): Vacuum cleaner, deciding to clean under your bed.

(D): Thermostat, deciding to cool down your fridge.

Questionnaire Answers

Table-Driven Agents

function TABLE-DRIVEN-AGENT(*percept*) **returns** an action

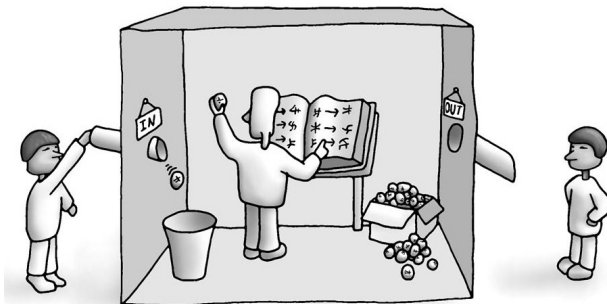
persistent: *percepts*, a sequence, initially empty

table, a table of actions, indexed by percept sequences, initially fully specified

append *percept* to the end of *percepts*

action \leftarrow LOOKUP(*percepts*, *table*)

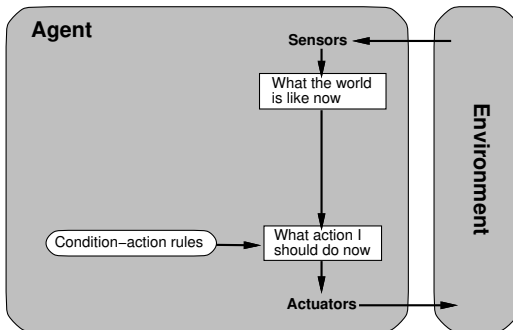
return *action*



Reflex Agents

More useful, but still very simple, method for choosing actions:

Condition-Action Rules: mappings between **current** percept and actions



→ Example?

What happens if the sensors do not always work?

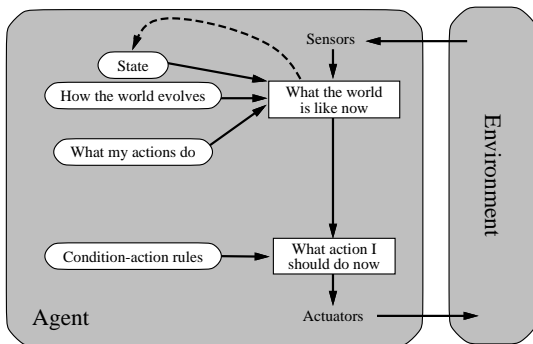
Reflex Model-based Agents

To handle failures in observing the environments, the agent needs to keep track of the part of the world it can't see now. **How?**

→ **Internal state**: depends on the percept history

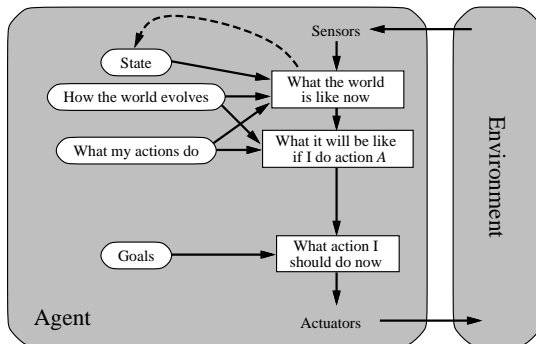
How to update the internal state? **Model of the world**:

- How the world evolves independently of the agent;
- How the agent's own actions affect the world.



Goal-Based Agents (Belief-Desire-Intention)

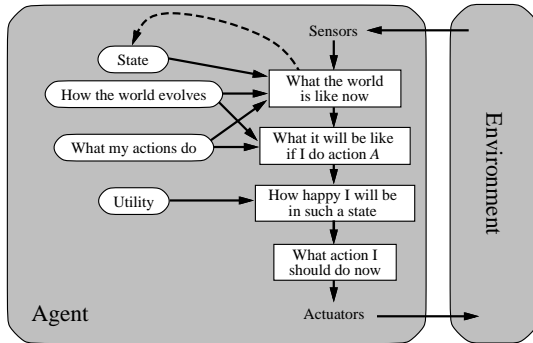
Often, doing the right thing requires considering the future:



→ Example?

Utility-Based Agents

Often, doing the right thing requires trading-off costs and benefits:



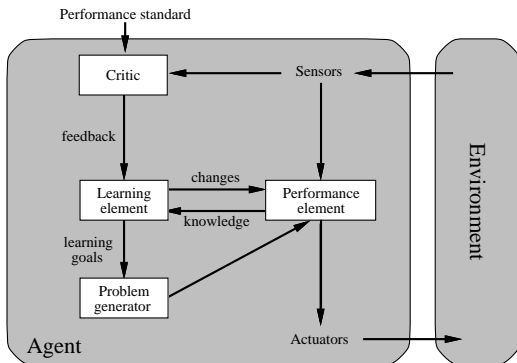
→ A **utility function** maps a state to a number which represents how desirable the state is.

Agent's utility function is an internalization of the performance function.

→ **Example?**

Learning Agents

It is typically useful to be able to **learn**: (extreme case: start with no knowledge in an unknown environment)



→ **Performance element**: Selects actions (**exploitation**). **Learning element**: Learns new knowledge. **Critic**: Measures performance. **Problem generator**: Suggests actions favoring informative learning experiences (**exploration**).

Domain-Specific vs. General Agents



Duell Kasparow gegen Deep Blue (1997): Demütigende Niederlage

vs.



vs.

vs.

Solver specific to a particular problem ("domain")

More efficient

Solver based on *description* in a general problem-description language (e.g., the rules of any board game)

More *intelligent*

Questionnaire

Question!

What kind of agent are you?

(A): Table-Driven

(B): Reflex Agent

(C): Learning

(D): Domain-Specific

The Environment of Rational Agents

- Fully observable vs. partially observable (Accessible vs. inaccessible)
Are the relevant aspects of the environment accessible to the sensors?
- Deterministic vs. stochastic
Is the next state of the environment completely determined by the current state and the selected action?
Stochastic: uncertainty is quantified by using probabilities.
Nondeterministic: uncertainty as actions with multiple outcomes.
Strategic: If the only non-determinism are actions of other agents.
- Episodic vs. sequential
Can the quality of an action be evaluated within an episode (perception + action), or are future developments decisive?

The Environment of Rational Agents, ctd.

- **Static** vs. **dynamic**

Can the environment change while the agent is deliberating?

If the environment does not change, but the agent's performance score changes, the environment is called **semi-dynamic**.

- **Discrete** vs. **continuous**

Is the environment discrete or continuous?

Distinction applies to time, environment states and agent's actions and percepts.

- **Single agent** vs. **multi-agent**

Is there just one agent, or several of them?

There are **competitive** and **cooperative** multi-agent scenarios.

Examples of Environments

Classifying AI Areas

Many sub-areas of AI can be classified by:

- Domain-specific vs. general.
- The environment.
- (Particular agent architectures sometimes also play a role, especially in Robotics.)

→ The same is true of the sub-topics in this course. The focus is on general methods (a bias in much of the AI field) and simple environments (after all, it's an introductory course only).

→ Up next: A rough classification of our topics, in these terms.

Classifying AI Areas: Our Topics

Classical Search

Environment:

- Fully observable vs. partially observable.
- Deterministic vs. stochastic.
- Episodic vs. sequential.
- Static vs. dynamic.
- Discrete vs. continuous.
- Single-agent vs. multi-agent.

Approach:

- Domain-specific vs. general.

Classifying AI Areas: Our Topics

Planning

Environment:

- Fully observable vs. partially observable.
- Deterministic vs. stochastic.
- Episodic vs. sequential.
- Static vs. dynamic.
- Discrete vs. continuous.
- Single-agent vs. multi-agent.

Approach:

- Domain-specific vs. general.

→ Planning formalisms and approaches exist also for any and all of partial observability, and stochastic/dynamic/continuous/multi-agent settings.

Classifying AI Areas: Our Topics

Adversarial Search

Environment:

- Fully observable vs. partially observable.
- Deterministic vs. stochastic.
- Episodic vs. sequential.
- Static vs. dynamic.
- Discrete vs. continuous.
- Single-agent vs. multi-agent.

Approach:

- Domain-specific vs. general.

→ Adversarial search formalisms and approaches exist also for partial observability and stochastic settings.

Classifying AI Areas: Our Topics

General Game Playing

Environment:

- Fully observable vs. partially observable.
- Deterministic vs. stochastic.
- Episodic vs. sequential.
- Static vs. dynamic.
- Discrete vs. continuous.
- Single-agent vs. multi-agent.

Approach:

- Domain-specific vs. general.

→ General game playing formalisms and approaches exist also for partial observability and stochastic settings.

Classifying AI Areas: Our Topics

Constraint Satisfaction & Reasoning

Environment:

- Fully observable vs. partially observable.
- Deterministic vs. stochastic.
- Episodic vs. sequential.
- Static vs. dynamic.
- Discrete vs. continuous.
- Single-agent vs. multi-agent.

Approach:

- Domain-specific vs. general.

Classifying AI Areas: Our Topics

Probabilistic Reasoning

Environment:

- Fully observable vs. **partially observable**.
- Deterministic vs. **stochastic**.
- **Episodic** vs. sequential.
- **Static** vs. dynamic.
- **Discrete** vs. continuous.
- **Single-agent** vs. multi-agent.

Approach:

- Domain-specific vs. **general**.

Questionnaire

Question!

James Bond's environment is?

(A): Fully Observable.

(B): Episodic.

(C): Static.

(D): Single-Agent.

Question!

Your own environment is?

(A): Fully Observable.

(B): Episodic.

(C): Static.

(D): Single-Agent.

Questionnaire Answers

Summary

- An **agent** is something that perceives and acts. It consists of an architecture and an agent program.
- A **rational agent** always takes the action that maximizes its expected performance, subject to the percept sequence and its environment knowledge.
- There are a variety of agent designs:
 - **Reflex agents** respond to percepts by condition-action rules.
 - **Reflex model-based agents** maintain a model of the world.
 - **Goal-based agents** work towards goals.
 - **Utility-based agents** make trade-offs using a utility function.
 - **Learning agents** improve their behavior over time.
- Some **environments** are more demanding than others ...
... your own, and that of James Bond, are the most difficult.

Reading

- *Chapter 2: Intelligent Agents* [Russell and Norvig (2020)].

Content: A much more detailed description of the subjects I overviewed here (agents, agent architectures, environments). Just as broad and superficial, though.

References I

Stuart Russell and Peter Norvig. *Artificial Intelligence: A Modern Approach (Fourth Edition)*. Prentice-Hall, Englewood Cliffs, NJ, 2020.