CS 5/7320 Artificial Intelligence

Intelligent Agents AIMA Chapter 2

Slides by Michael Hahsler based on slides by Svetlana Lazepnik with figures from the AIMA textbook.

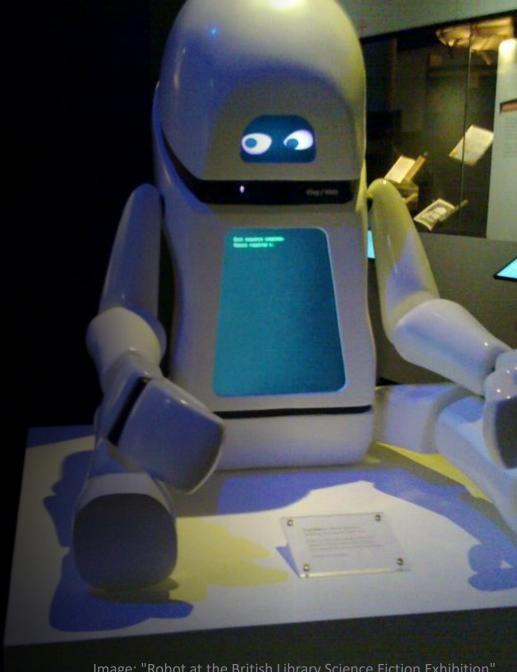
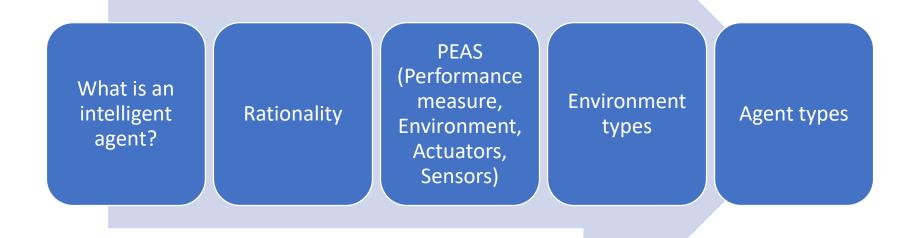


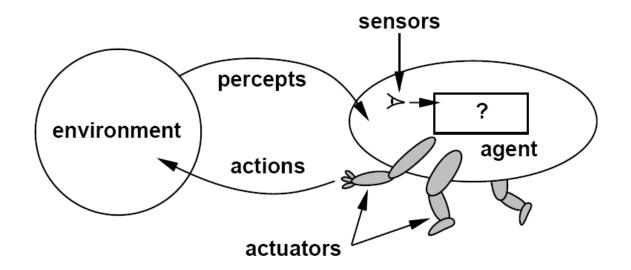
Image: "Robot at the British Library Science Fiction Exhibition" by BadgerGravling

Outline



What is an Agents?

 An agent is anything that can be viewed as perceiving its environment through sensors and acting upon that environment through actuators.



- Control theory: A closed-loop control system (= feedback control system)
 is a set of mechanical or electronic devices that automatically regulates a
 process variable to a desired state or set point without human interaction.
 The agent is called controller.
- **Softbot**: Agent is a software program that runs on a host device.

Agent function and agent program

The agent function maps from percept sequences P^* to actions A formulated as an abstract mathematical function $f: P^* \to A$ (e.g., a table).

The agent program is the concrete implementation running in a physical system.

Agent = architecture (hardware) + agent program



- Sensors
- Memory
- Computational power

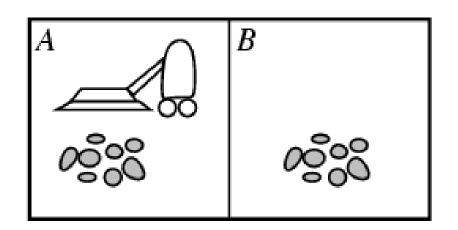
Example: Vacuum-cleaner world

Percepts:

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

Actions:

Left, Right, Suck, NoOp



Last Percept

Agent function: $f: P^* \rightarrow A$

Percept Sequence	Action
[A, Clean]	Right
[A, Dirty]	Suck
 [A Clean] [B Clean]	l eft

[A, Clean], [B, Clean], [A, Dirty] Suck

Implemented agent program:

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
```

This table can become infinitively large!

Rational agents: What is good behavior?

Foundation

- **Consequentialism**: Evaluate behavior by its consequences.
- Utilitarianism: maximize happiness and well-being

Definition of a 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 the **agent's built-in knowledge**."

- Performance measure: An *objective* criterion for success of an agent's behavior (often called utility function).
- Expectation: Outcome averaged over all possible situations that may arise.

This means:

- Rationality ≠ Omniscience (rational agents can make mistakes if percepts and knowledge do not suffice to make a good decision)
- Rationality ≠ Perfection (rational agents maximize expected outcomes not actual outcomes)
- It is rational to explore and learn (i.e., use percepts to supplement prior knowledge and become autonomous)

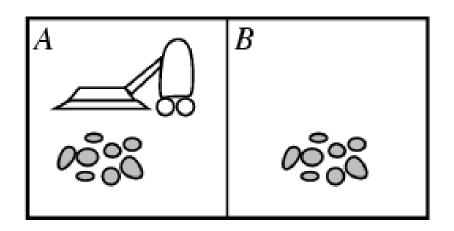
Example: Vacuum-cleaner world

Percepts:

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Actions:

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Agent function:	
Percept Sequence [A, Clean] [A, Dirty]	Action Right Suck
 [A, Clean], [B, Clean] 	Left

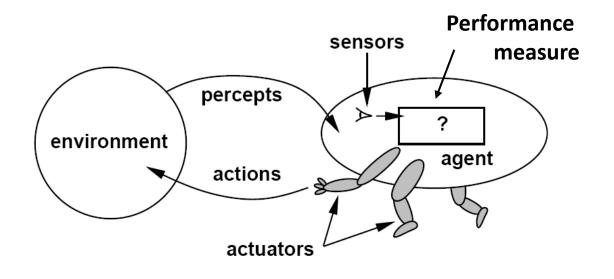
```
Implemented agent program:

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
```

What could be a performance measure? Is this agent rational?

Problem specification: PEAS



Performance measure

Environment

Actuators

Sensors

Example: 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

Example: Spam filter

Performance measure

 Minimizing false positives, false negatives

Environment

- A user's email account
- email server

Actuators

- Mark as spam
- delete
- etc.

Sensors

- Incoming messages
- other information about user's account

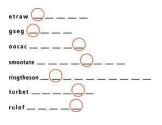
Environment types

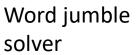
- Fully observable (vs. partially observable): The agent's sensors give it access to the complete state of the environment at each point in time.
- Deterministic (vs. stochastic): The next state of the environment is completely determined by the current state and the agent's action.
 - Strategic: the environment is deterministic except for the actions of other agents are stochastic but follow a strategy.
- Episodic (vs. sequential): Episode = get precept + do action.
 The agent's choice of action in one episode does not affect the next episodes.

Environment types

- Static (vs. dynamic): The environment is unchanged while an agent is deliberating.
 - Semidynamic: the environment does not change with the passage of time, but the agent's performance score does.
- Discrete (vs. continuous): The environment provides a fixed number of distinct percepts, actions, and environment states.
 - Time can also evolve in a discrete or continuous fashion
- Single agent (vs. multi-agent): An agent operating by itself in an environment.
- Known (vs. unknown): The agent knows the rules of the environment.

Examples of different environments







Chess with a clock



Scrabble



Taxi driving

Observable	Fully	Fully	Partially	Partially
Deterministic	Deterministic	Strategic	Stochastic +Strategic	Stochastic
Episodic	Episodic	Sequential	Sequential	Sequential
Static	Static	Semidynamic	Static	Dynamic
Discrete	Discrete	Discrete	Discrete	Continuous
Single agent	Single	Multi	Multi	Multi Actio

Actions have long-term effects.

Hierarchy of agent types

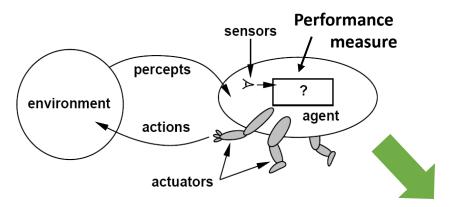
Utility-based agents

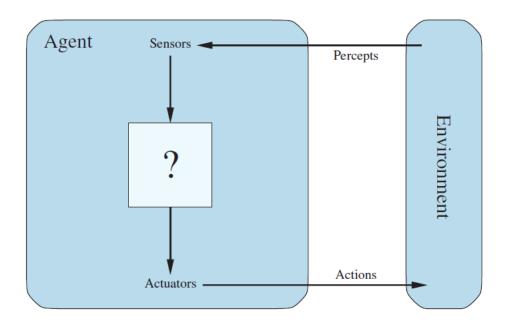
Goal-based agents

Model-based reflex agents

Simple reflex agents

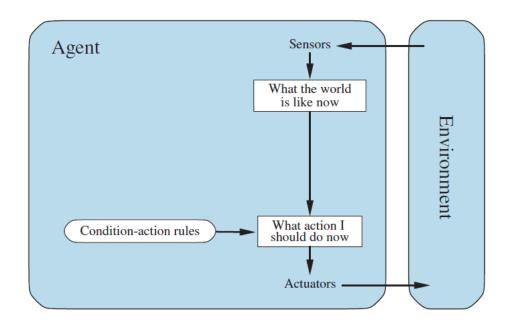
Structure of an Agent





Simple reflex agent

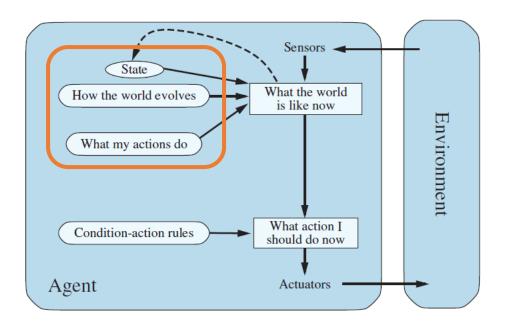
Rules select action only **based on current percept**, ignoring all past percepts (no memory). This is typically very fast!



Example: simple, rule-based vacuum cleaner from before.

Model-based reflex agent

Maintains internal state (memory) to keeps track of aspects of the environment that cannot be currently observed. There is now more information for the rules to making decisions.

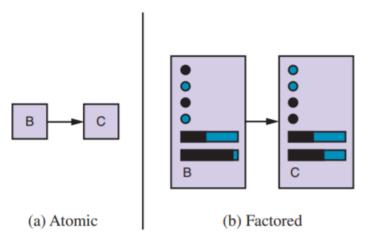


Example: simple, vacuum cleaner that remembers were it has already cleaned.

State representation

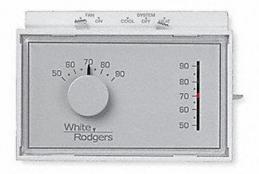
States help to keep track of the environment. The representation can be

- Atomic: Just a label for a black box. E.g., A, B
- **Factored**: A vector of attribute values. E.g., [location = left, status = clean, temperature = 75 deg. F]



State Space: The set of all possible states.

Old-school vs. Smart thermostat



Old-school thermostat

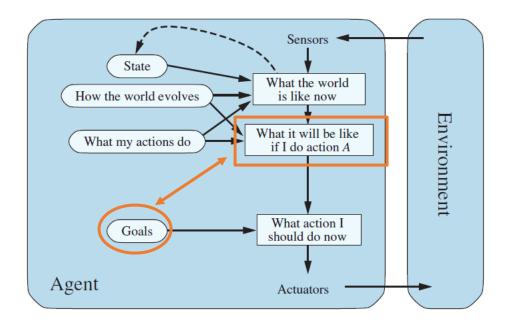
Percepts States



Percepts States

Goal-based agent

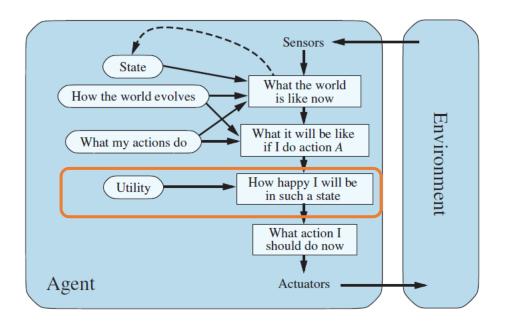
The agent chooses actions in the current state to reach a **goal state** as fast as possible. We need **search algorithms** to find action sequences that reach that goal.



Example: Solving a puzzle. What action gets me closer to the solution?

Utility-based agent

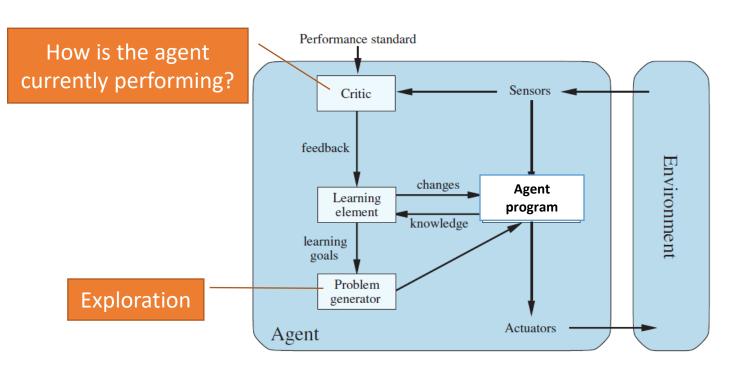
The agent uses a utility function to evaluate the **desirability of different states**. Choose the action to maximize expected utility over time (i.e., stay in desirable states).



Example: An autonomous Mars rover prefers states where its battery is not critically low.

Agents that Learn

The **learning element** modifies the agent program (reflex-based, goal-based, or utility-based) to improve its performance.



Smart thermostat



Change temperature when you are too cold/warm.

Reflex Agent? Goalibased **Utility-based?**

Smart thermostat

Percepts

- Temp: deg. F
- Outside temp.
- Weather report
- Energy curtailment
- Someone walking by
- Someone changes temp.
- Day & time
- ...

States

Factored states

- Estimated time to cool the house
- Someone home?
- How long till someone is coming home?
- A/C: on, off

What type of intelligent agent is this?

Features are:

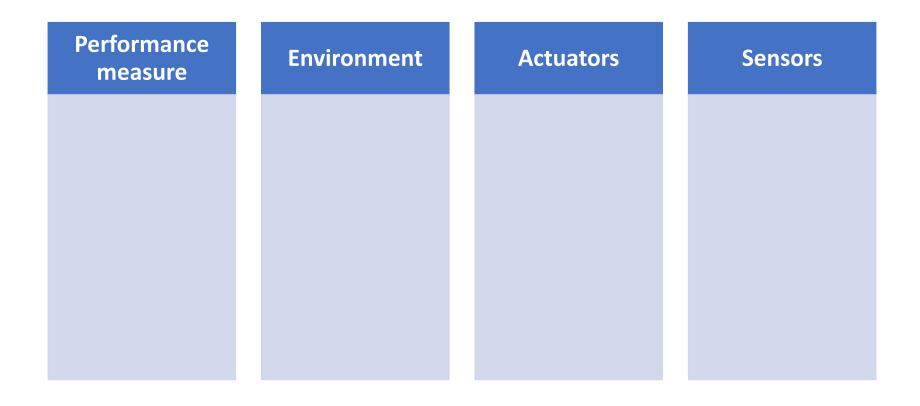
- Control via App
- Cleaning Modes
- Navigation
- Mapping
- Boundary blockers



iRobot's Roomba brand has become as synonymous with robot vacuum as Q-tips is with cotton swabs. The Wi-Fi-enabled Roomba 960 is ample evidence why. It turns a tiresome chore into something you can almost look forward to. With three cleaning modes and dirt-detecting sensors, it kept all the floor surfaces in our testing immaculate, and its camera-driven navigation and mapping were superb. Its easy-to-use app provides alerts and detailed cleaning reports. The ability to control it with Amazon Alexa and Google Home voice commands are just the cherry on top.

Source: https://www.techhive.com/article/3269782/best-robot-vacuum-cleaners.html

PEAS Description of a modern robot vacuum



What type of intelligent agent is a modern robot vacuum?

How would the utility be **Utility-based agents** defined? ls it learning? Goal-based agents Does it have a goal state? Does it use states. How Model-based reflex agents would they be defined (atomic/factored)? Simple reflex agents Does it use simple reflexes?

Conclusion

Intelligent agents inspire the research areas of modern Al

Search for a goal (e.g., navigation).

Optimize functions (e.g., utility).

Stay within given constraints

(constraint satisfaction problem; e.g., reach the goal without running out of power)

Deal with **uncertainty** (e.g., current traffic on the road).

Learn a good agent program from data and improve over time (machine learning).

Sensing

(e.g, natural language processing, vision)