



COMMONWEALTH OF AUSTRALIA

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FIT5047: Fundamentals of AI

Intelligent Agents Chapter 2

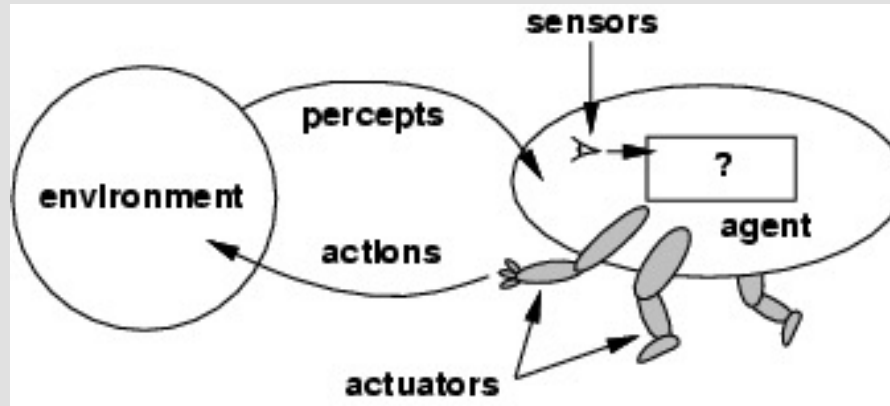
Outline

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

Agents

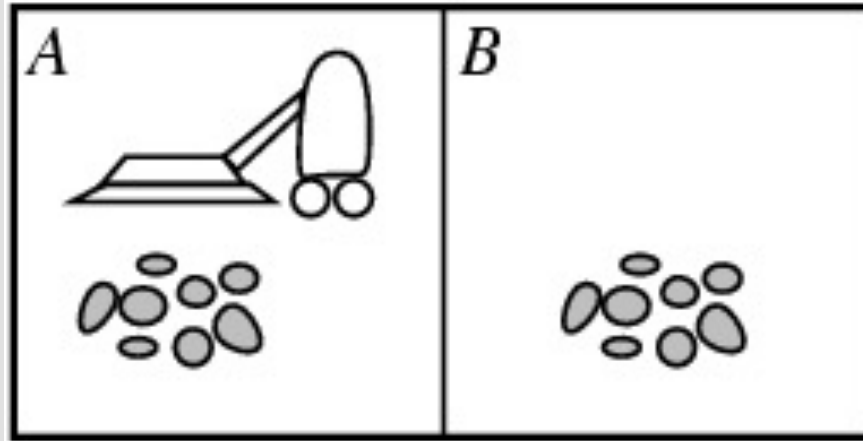
- An **agent** is anything that can be viewed as **perceiving its environment using sensors, and acting upon that environment via actuators**
- **Human agent:**
 - eyes, ears and other organs for sensors
 - hands, legs, mouth and other body parts for actuators
- **Robotic agent:**
 - cameras and infrared 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**

Example: Agent in a vacuum-cleaner world



- **Percepts:** location and contents, e.g., [A,Dirty]
- **Actions:** *Left, Right, Suck*
- **Program:**
 If *status=Dirty* **return** Suck
 Elseif *Location=A* **return** Right
 Elseif *Location=B* **return** Left

**Is this a
rational agent?**



Rationality and rational agents

- **Rationality depends on**

- Performance measure
- The agent's prior knowledge of the environment
- The actions that the agent can perform
- The percept sequence to date

- **Definition:**

*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***

Rational autonomous agents

- Rationality is NOT omniscience
- Agents can perform actions to modify future percepts in order to obtain useful information
→ **exploration, learning**
- An agent is **autonomous** if its behavior is determined by its own experience

Task environment – PEAS

To design a rational agent, we must specify the Task environment

- **PEAS**
 - Performance measure
 - Environment
 - Actuators
 - Sensors

PEAS – Example (I)

Automated taxi driver:

- **Performance measure**
 - Safe, fast, legal, comfortable trip, minimize fuel consumption, maximize profit
- **Environment**
 - Road types, road contents, customers, operating conditions
- **Actuators**
 - Control over the car, interfaces for informing other vehicles and informing passengers
- **Sensors**
 - Cameras, sonar, speedometer, GPS, odometer, engine sensors, interface for receiving information from other vehicles and passengers (e.g., speech recognizer)



PEAS – Example (II)

Internet shopping agent:

- **Performance measure**
 - cheap, good quality, appropriate product
- **Environment**
 - current internet sites, vendors
- **Actuators**
 - Packages that display to user, follow URL, fill in forms
- **Sensors**
 - Packages that read HTML pages (text, graphics, scripts)

Environment types (I)

The environment type largely determines the agent design

- **Fully (partially) observable** – An agent's sensors give it access to the complete state of the environment at all times
- **Known (unknown)** – An agent knows the “laws” of the environment
- **Single (multi) agent** – An agent operating by itself in an environment
- **Deterministic (stochastic)** – The next state is completely determined by the current state and the action executed by the agent

Environment types (II)

- **Episodic (sequential)** – The agent's experience is divided into atomic *episodes*. The next episode does **NOT** depend on previous actions
 - In each episode an agent perceives a percept and performs a single action
- **Static (dynamic)** – The environment is unchanged while an agent is deliberating
- **Discrete (continuous)** – Pertains to number of states, the way time is handled, and number of percepts and actions
 - E.g., states may be continuous, but actions be discrete

Environment types – Examples

	Sorting laundry	8-puzzle	Back-gammon	Medical diagnosis	Taxi
Observable?					
Known?					
Single agent?					
Deterministic?					
Episodic?					
Static?					
Discrete?					

The real world is partially observable, unknown, multi-agent, stochastic, sequential, dynamic, continuous

Environments and methodologies

	Search	Logical inference	Bayesian networks	Machine learning
Observable?	✓	✓		
Known?	✓	✓	✓	x
Single agent?				
Deterministic?	✓	✓	x	
Episodic?				
Static?	✓	✓	✓	✓
Discrete?	✓	✓		

Agent functions and programs

- An agent is specified by its **agent function** which maps percept sequences to actions
- **Aim**: design a program that implements the rational agent function concisely

Agent types

Based on the function = how actions are selected

Agent type	Action selected based on
Simple reflex	current percept
Model based	+ internal state (world model)
Goal based	+ goal
Utility based	+ utility function
Learning	performance element = above agent + critic + learning element + problem generator (exploratory)



Agent types: Taxi example

Agent type	Action
Simple reflex	brake when brake-lights of car in front light up
Model based	+ remember the roads travelled, time, state
Goal based	+ make a plan to reach a destination
Utility based	+ quickest with least petrol consumption
Learning performance elem +critic +learning element +problem generator	above agent observes the world & informs learning elem formulates new driving rules based on the feedback from critic + perf agent knowledge might suggest some driving exercises



How components of agent programs work?

Depends on the representation of states:

- **Atomic** – each state is indivisible
(Search, Game playing, Markov Decision Processes)
- **Factored** – splits each state into attributes, each of which has a value (Propositional logic, Bayesian networks, Machine learning)
- **Structured** – represents how things are related to each other (First-order logic, First-order probability models)

A more expressive representation can capture everything a less expressive representation can capture, but reasoning and learning is harder

Reading

- **Russell, S. and Norvig, P. (2010),
Artificial Intelligence – Chapter 2**

Next topic

- **Lecture Topic 3 (LN3)**
 - Problem solving as search