MONASH University Information Technology



COMMONWEALTH OF AUSTRALIA

Copyright Regulations 1969

WARNING

This material has been reproduced and communicated to you by or on behalf of Monash University pursuant to Part VB of the Copyright Act 1968 (the Act).

The material in this communication may be subject to copyright under the Act. Any further reproduction or communication of this material by you may be the subject of copyright protection under the Act.

Do not remove this notice.





FIT5047 – Intelligent Systems

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 through sensors, and acting upon that environment through actuators

Human agent:

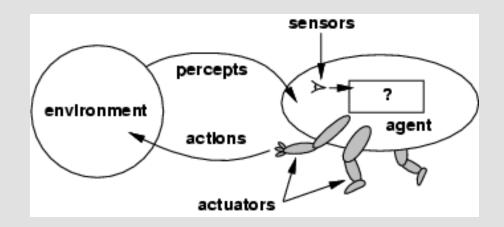
- eyes, ears and other organs for sensors
- hands, legs, mouth and other body parts for actuators

Robotic agent:

- cameras and infrared range finders for sensors
- various motors for actuators



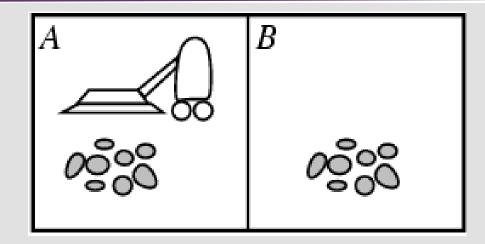
Agents and Environments



- The agent function maps from percept histories to actions:
 f: P* → A
- The agent program runs on the physical architecture to produce f
- agent = architecture + program



Example: Vacuum-cleaner World and Agent



- Percepts: location and contents, e.g., [A,Dirty]
- Actions: Left, Right, Suck
- Program:

If status=Dirty return Suck
Elself Location=A return Right
Elself 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 <u>rational</u> <u>agent</u> 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 WWW sites, vendors
- Actuators
 - display to user, follow URL, fill in form
- Sensors
 - 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., state may be continuous, but actions may 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, multiagent, stochastic, sequential, dynamic, continuous



Environments and Methodologies

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



Agent Functions and Programs

- An agent is completely specified by the <u>agent</u> <u>function</u> that 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	<pre>performance element = above agent + critic + learning element + problem generator (exploratory)</pre>



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 the critic might suggest some driving exercises



How Components of Agent Programs Work?

Depends on the representations of states:

- Atomic each state is indivisible (Search, Game playing)
- 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)



Reading

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



Next Lecture Topic

- Lecture Topic 3
 - Problem Solving as Search

