COMP9414/9814/3411 18s1 Agents

Agent Types

In this course we will consider five different types of agent:

- Reactive Agent
- Model-Based Agent
- Planning Agent
- Game Playing Agent
- Learning Agent

UNSW © Alan Blair, 2013-18

© Alan Blair, 2013-18

COMP9414/9814/3411 18s1

UNSW

Agents

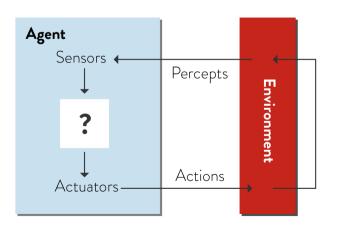
2

COMP9414/9814/3411 18s1

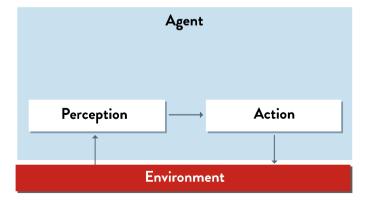
Agents

3

Agent Model



Reactive Agent



UNSW © Alan Blair, 2013-18 UNSW

© Alan Blair, 2013-18

COMP9414/9814/3411 18s1 Agents 4 COMP

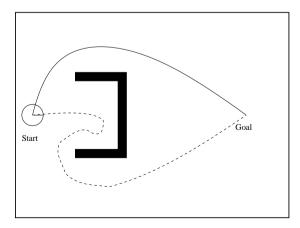
Reactive Agents

- Choose the next action based only on what they currently perceive, using a "policy" or set of rules which are simple to apply
- Sometimes pajoratively called "simple reflex agents" but they can do surprisingly sophisticated things!
 - Swiss robots
 - simulated hockey

UNSW © Alan Blair, 2013-18

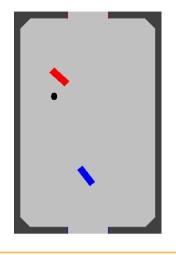
COMP9414/9814/3411 18s1 Agents

Limitations of Reactive Agents



COMP9414/9814/3411 18s1 Agents 5

Reactive Agents





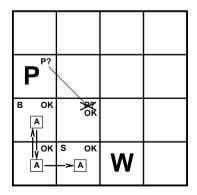
UNSW © Alan Blair, 2013-18

COMP9414/9814/3411 18s1 Ag

UNSW

Agents

Limitations of Reactive Agents



Sometimes we need to remember previous percepts in order to make sensible choices.

11

Limitations of Reactive Agents

- Reactive Agents have no memory or "state"
 - unable to base decision on previous observations
 - ▶ may repeat the same sequence of actions over and over

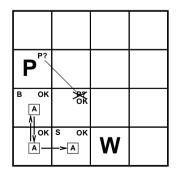
Agents

- This phenomenon can also be observed in nature
 - wasp dragging stung grasshopper into its nest

UNSW © Alan Blair, 2013-18

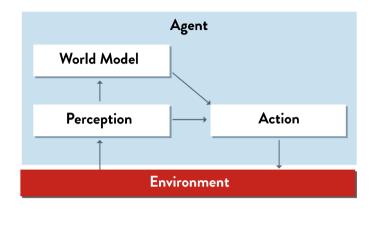
COMP9414/9814/3411 18s1 Agents

Advantages of Model-Based Agent



■ Model-Based Agent can keep a "map" of the places it has visited, and remember what it perceived there.

Model-Based Agent



UNSW © Alan Blair, 2013-18

COMP9414/9814/3411 18s1 Agents

Limitations of Model-Based Agents

?	S	?	?
?	В	?	В
В		В	

Sometimes we may need to plan several steps into the future.

10

COMP9414/9814/3411 18s1 Agents 12 COMP9414/9814/3411 18s1 Agents 13

Limitations of Model-Based Agent

An agent with a world model but no planning can look into the past, but not into the future; it will perform poorly when the task requires any of the following:

- searching several moves ahead
 - ► Chess, Rubik's cube
- complex tasks requiring many individual steps
 - ▶ cooking a meal, assembling a watch
- logical reasoning to achieve goals
 - travel to New York

UNSW

PLANNING AGENT

- Decision making of this kind is fundamentally different from the condition action rules.
- It involves consideration of the future
- 1. "What will happen if I do such and such?" and
- 2. "Will that make me happy?"

In the reflex agent designs, this information is not explicitly represented, because the built-in rules map directly from

Agent

World Model

Planning

Perception

Action

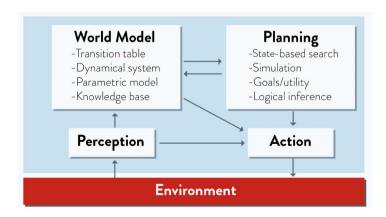
Environment

© Alan Blair, 2013-18 UNSW © Alan Blair, 2013-18

Planning Agent

COMP9414/9814/3411 18s1 Agents 14 COMP9414/9814/3411 18s1 Agents 15

Models and Planning



Reasoning about Future States

?	?	?	?
?	?	?	?
?	?	?	?
∧ S	?	?	?

What is the best action in this situation?

UNSW © Alan Blair, 2013-18

UNSW

COMP9414/9814/3411 18s1 Agents 16 COMP9414/9814/3411 18s1 Agents

18

Faking it

■ Sometimes an agent may appear to be planning ahead but is actually just applying reative rules.

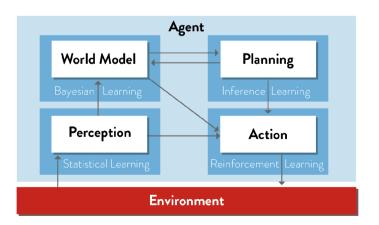
```
if( Glitter ) then
        Grab
else if( Stench ) then
        Shoot
else
        randomly Left, Right or Forward
```

- These rules can be hand-coded, or learned from experience.
- Agent may appear intelligent, but is not flexible in adapting to new situations.

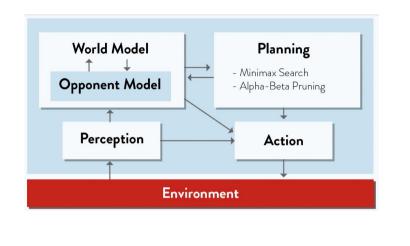
UNSW © Alan Blair, 2013-18

COMP9414/9814/3411 18s1 Agents

Learning Agent



Game Playing Agent



COMP9414/9814/3411 18s1

UNSW

Agents

19

© Alan Blair, 2013-18

Learning

- Learning is not a separate module, but rather a set of techniques for improving the existing modules
- Learning is necessary because:
 - ▶ may be difficult or even impossible for a human to design all aspects of the system by hand
 - ▶ the agent may need to adapt to new situations without being re-programmed by a human

UNSW © Alan Blair, 2013-18

UNSW

© Alan Blair, 2013-18

COMP9414/9814/3411 18s1 Agents 20

Learning

We must distinguish complexity of learning from complexity of application.

The policy for the simulated hockey player took several days of computation to derive (in this case, by evolutionary computation) but, once derived, it can be applied in real time.

UNSW © Alan Blair, 2013-18