

# Representational Dimensions

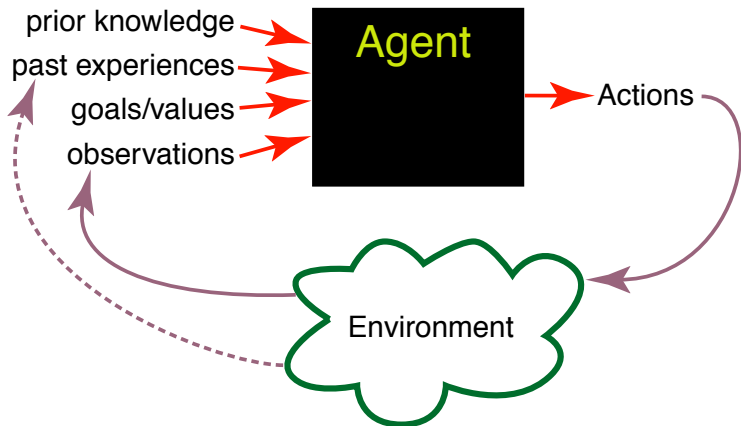
CPSC 322 Lecture 2

September 12, 2007

# Essentials

- **Course web site:** <http://www.ugrad.cs.ubc.ca/~cs322>
  - This is where most handouts and slides will be distributed
- **Textbook:** *Computational Intelligence, 2nd Edition*, by Poole, Mackworth and Goebel. Still under development; electronic version only.
- **WebCT:** used for textbook, discussion board, solutions to assignments
  - Use the discussion board for questions about assignments, material covered in lecture, etc, rather than email
  - Use email for private questions (e.g., health problems)
- You get 4 **late days** for assignments
- **Assignment 0** is due on Monday, no late days
  - Use handin to submit it, and use WebCT to learn about handin.

# Agents acting in an environment



# Representation

It turns out that when we want to think clearly and precisely about action, **representation** is critical:

- What different **configurations** can the world be in, and how do we denote them in a computer?
- What sorts of **beliefs** can we have about what configuration the world is in, and are these beliefs certain?
- How would the world be changed if we were to take some given action: what are the **system dynamics**?

# What do we want from a representation?

## We want a representation to be:

- **rich enough** to express the knowledge needed to solve the problem.
- as **close to the problem** as possible: compact, natural and maintainable.
- amenable to **efficient computation**; able to express features of the problem we can exploit for computational gain.
- **learnable** from data and past experiences.
- able to **trade off** accuracy and computation time.

# Representation and Reasoning System

Problem  $\implies$  representation  $\implies$  computation

A **representation and reasoning system** (RRS) consists of

- A **language** in which a model of the world can be described symbolically (“syntax”)
- A way to assign **meaning** to the symbols (“semantics”)
- Computational **procedures** to compute answers or solve problems

Example RRSs:

- Programming languages: Fortran, C++,...
- Natural Language

This course will explore alternatives that fall **between these extremes**.

# Overview of this course

**This course will emphasize two main themes:**

## Reasoning

How should an agent **act** given the current state of its environment and its goals?

## Representation

How should the environment **be represented** in order to help an agent to reason effectively?

# Representations considered this course

**Furthermore, the course will consider three main representational dimensions:**

- 1 Deterministic vs. stochastic domains
- 2 Static vs. sequential domains
- 3 Single-agent vs. multiagent domains



# 1. Deterministic vs. Stochastic Domains

Historically, AI has been divided into two camps: those who prefer representations based on logic and those who prefer probability.

- Is the environment **deterministic** or **stochastic**?
- Is the agent's knowledge **certain** or **uncertain**?

A few years ago, CPSC 322 covered logic, while CPSC 422 introduced probability

- now we introduce both representational families in 322, and 422 goes into more depth
- this should give you a better idea of what's included in AI

Some of the most exciting current research in AI is actually building bridges between these camps.

## 2. Static vs. Sequential Domains

### How many actions does the agent need to select?

- The agent needs to take a **single action**
  - solve a Sudoku
  - diagnose a patient with a disease
- The agent needs to take a **sequence of actions**
  - navigate through an environment to reach a goal state
  - bid in online auctions to purchase a desired good
- Important caveat:
  - in deterministic domains, the **distinction** between static and sequential settings may seem somewhat artificial
    - we can redefine actions (e.g., fill in individual numbers in the Sudoku vs. solving the whole thing)
    - indeed, some of the same techniques work in both settings
  - the same cannot be said about **stochastic domains**

### 3. Single-agent vs. Multiagent domains

One final representational question is whether the environment **includes other agents**

- Everything we've said so far presumes that there is only one agent in the environment.
- If there are other agents whose actions affect us, it can be useful to **explicitly model their goals and beliefs** rather than considering them to be part of the environment
- Agents can have their own goals: cooperative, competitive, or a bit of both

We'll end the course by looking briefly at **multiagent systems**

# Modules we'll cover in this course

- 1 Making single and sequential decisions in deterministic environments
  - Search, CSPs, Planning
- 2 Richer representations in deterministic environments:
  - Logic
- 3 Making single decisions in stochastic environments:
  - Bayes Nets, Influence Diagrams
- 4 Making sequential decisions in stochastic environments:
  - Influence Diagrams, MDPs
- 5 Multiagent systems (*if time permits*)
  - Zero-sum games; Nash equilibria

# Dimensions of Representational Complexity

## **We've already discussed:**

- Deterministic versus stochastic domains
- Static versus sequential domains
- Single-agent versus multiagent domains

## **Some other important dimensions of complexity:**

- Explicit state or propositions or relations
- Flat or hierarchical
- Knowledge given versus knowledge learned from experience
- Goals versus complex preferences
- Perfect rationality versus bounded rationality

# Explicit State or propositions or relations

## How do we model the world?

- You can enumerate the **states** of the world.
- A state can be described in terms of **features**.
  - Often it is more natural to describe states in terms of assignments of values to variables (features).
    - 30 binary features can represent  $2^{30} = 1,073,741,824$  states.
- Features can be described in terms of **objects** and **relationships**.
  - There is a feature for each relationship on each tuple of individuals.
  - One binary relation and 10 individuals can represent  $10^2 = 100$  propositions and  $2^{100}$  states.

# Flat or hierarchical

**Is it useful to model the whole world at the same level of abstraction?**

- You can model the system at one level of abstraction: **flat**
- You can model the system at multiple levels of abstraction: **hierarchical**
- **Example:** Planning a trip from here to a resort in Cancun, Mexico

# Knowledge given versus knowledge learned from experience

## How much do we know about the world in advance?

- The agent is provided with a model of the world before it starts to act
- The agent must learn how the world works based on experience
  - in this case, the agent often still does start out with some **prior knowledge**



# Goals versus complex preferences

**If an agent doesn't want to achieve anything, it has no reason to act. How do we represent an agent's desire(s)?**

- An agent may have a **goal** that it wants to achieve
  - e.g., there is some state or set of states of the world that the agent wants to be in
  - e.g., there is some proposition or set of propositions that the agent wants to make true
- An agent may have complex **preferences**
  - e.g., there is some preference function that describes how happy the agent is in each state of the world; the agent's task is to put the world into a state which makes it as happy as possible

# Perfect rationality versus bounded rationality

**We've defined rationality as an abstract ideal. Is the agent able to live up to this ideal?**

- **Perfect rationality:** the agent can derive what the best course of action is.
- **Bounded rationality:** the agent must make good decisions based on its perceptual, computational and memory limitations.

# Modules we'll cover in this course

- 1 Making single and sequential decisions in deterministic environments
  - Search, CSPs, Planning
- 2 Richer representations in deterministic environments:
  - Logic
- 3 Making single decisions in stochastic environments:
  - Bayes Nets, Influence Diagrams
- 4 Making sequential decisions in stochastic environments:
  - Influence Diagrams, MDPs
- 5 Multiagent systems (*if time permits*)
  - Zero-sum games; Nash equilibria