Michael Royatsos mrovatso@inf.ed.ac.uk

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Lecture 11 - Agent Architectures 18th February 2005

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Where are we?

Last time

- Introduction to agents and multiagent systems
- Discussed key properties of agents (autonomy, rationality, social ability)
- Looked at different kinds of interaction
 - (coordination, communication, collaboration etc.)
- Discussion of key research topics in agents

Today ...

Agent Architectures

Symbolic AI: A Critical View

- ▶ Recall first lecture: symbol system vs. physical grounding hypothesis
 - Is inference on symbols representing the world sufficient to solve real-world problems . . .
 - ... or are these symbolic representations irrelevant as long as
 - the agent is successful in the physical world? "Elephants don't play chess" (or do they?)
- Also problems with "symbolic Al":
 - Computational complexity of reasoning in real-world applications
 - The knowledge acquisition bottleneck
 - Largely focuses on theoretical reasoning about the world

Types of Agent Architectures

- From this dispute a distinction between reactive (often called behaviour-based) and deliberative agents evolved
- Alternative view: distinction arises naturally from tension between reactivity and proactiveness (see previous lecture)
- Broad categories:
 - Deliberative Architectures focus on planning and symbolic reasoning
 - Reactive Architectures
 - focus on reactivity based on behavioural rules Hybrid Architectures
 - attempting to balance proactiveness with reactivity

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The RDI Architecture

- BDI: Beliefs, Desires, Intentions
- Based on work on human practical reasoning, i.e. everyday reasoning about "what to do"

Practical reasoning is a matter of weighing conflicting considerations for and against competing options, where the relevant considerations are provided by what the agent desires/values/cares about and what the agent believes. (Michael Bratman)

 Theoretical reasoning is rather directed towards beliefs and knowledge and usually involves no activity

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Intentions

- Bratman's model suggests the following properties:
 - Intentions pose problems for agents, who need to determine ways of achieving them
 - Intentions provide a 'filter' for adopting other intentions, which must not conflict
 - Agents track the success of their intentions, and are inclined to try again if their attempts fail
 - · Agents believe their intentions are possible
 - Agents do not believe they will not bring about their intentions
 Under certain circumstances, agents believe they will bring
 - Under certain circumstances, agents beli about their intentions
 - Agents need not intend all the expected side effects of their intentions

Practical Reasoning

- Practical reasoning consists of two main activities:
 - 1. Deliberation
 - 2. Means-ends reasoning

Combining these appropriately is the foundation of deliberative agency

- Deliberation is concerned with determining what one wants to achieve (considering one's preferences, choosing goals to pursue, etc.)
- Deliberation generates intentions
- Means-ends reasoning is used to determine how the goals are to be achieved (thinking about suitable actions, resources and how to "organise" activity)
- ► Means-ends reasoning generates plans

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Practical Reasoning Systems

Intentions

- Cohen-Levesque theory of intentions based on notion of persistent goal
- An agent has a persistent goal of φ iff:
 - 1. It has a goal that ϕ eventually becomes true, and believes that ϕ is not currently true
 - Before it drops the goal φ, one of the following conditions must hold:
 - the agent believes φ has been satisfied
 - the agent believes φ will never be satisfied
- Definition of intention (consistent with Bratman's list):

An agent intends to do action α iff it has a persistent goal to have brought about a state wherein it believed it was about to do α , and then did α .

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Desires

- Desires describe the states of affairs that are considered for achievement, i.e. basic preferences of the agent
- Desires are much weaker than intentions, they are not directly related to activity:

My desire to play basketball this afternoon is merely a potential influencer of my conduct this afternoon. It must vie with my other relevant desires [. . .] before it is settled what I will do. In contrast, once I intend to play basketball this afternoon, the matter is settled: I normally need not continue to weigh the pros and cons. When the afternoon arrives, I will normally just proceed to execute

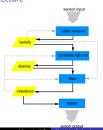
my intentions. (Bratman, 1990) Informatics UoE Knowledge Engineering

The BDI Architecture

Sub-components of overall BDI control flow:

- Belief revision function
 - Update beliefs with sensory input and previous belief
- Generate options
 - Use beliefs and existing intentions to generate a set of alternatives/options (=desires)
- Filtering function
 - · Choose between competing alternatives and commit to their achievement
- Planning function
 - Given current belief and intentions generate a plan for action
- Action generation: iteratively execute actions in plan sequence (in very simple model)

The RDI Architecture



Issues

- Different commitment strategies:
 - Blind/fanatical commitment: maintain intention until it has been achieved
 - Single-minded commitment: maintain intention until achieved or proves impossible
- Commitment both to ends (intention) and means (plan), particular commitment strategy may lead to overcommitment
- Re-planning: include a test for viability of plan after every action (and plan again)
- Intention reconsideration
 - Stop to think whether intentions are already
 - fulfilled/impossible to achieve
 - Trade-off: intention reconsideration is costly but necessary
 - meta-level control might be useful
 - Reconsideration always successful if agent would have changed intentions had he deliberated again

▶ BDI certainly most widespread model of rational agency, but also criticism as it is based on symbolic AI methods

- Some of the (unsolved/insoluble) problems of symbolic Al have lead to research in reactive architectures
- ► One of the most vocal critics of symbolic AI: Rodney Brooks
- ▶ Brooks has put forward three theses:
 - Intelligent behaviour can be generated without explicit representations of the kind that symbolic Al proposes
 - Intelligent behaviour can be generated without explicit abstract reasoning of the kind that symbolic AI proposes
 - Intelligence is an emergent property of certain complex systems

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Introduction re Architectures

Subsumption Architecture

Subsumption Architect Discussion

Example

- ▶ Luc Steels' cooperative mars explorer system
- Domain: a set of robots are attempting to gather rock samples on Mars (location of rocks unknown but they usually come in clusters); there is a radio signal from the mother ship to find way back
- Only five rules (from bottom (high priority) to top (low priority)):
 - 1. If detect an obstacle then change direction
 - 2. If carrying samples and at the base then drop samples
 - If carrying samples and not at the base then travel up signal gradient
 - 4. If detect a sample then pick sample up
- If true then move randomly
- Near-optimal behaviour!

Subsumption Architecture

- Brooks' research based on two key ideas:
 - Situatedness/embodiment: Real intelligence is situated in the world, not in disembodied systems such as theorem provers or expert systems
 - Intelligence and emergence: Intelligent behaviour result from agent's interaction with its environment. Also, intelligence is "in the eye of the beholder" (not an innate property)
- Subsumption architecture illustrates these principles:
 - Essentially a hierarchy of task-accomplishing behaviours (simple rules) competing for control over agent's behaviour
 - Lower layers correspond to "primitive" behaviours and have precedence over higher (more abstract) ones
 - Extremely simple in computational terms (but sometimes extremely effective)

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Discussion

- Reactive architectures achieve tasks that would be considered very impressive using symbolic AI methods
- But also some drawbacks:
 - If it works, how do we know why it works?
 - departure from "knowledge level" implies of transparency
 What if it doesn't work?
 - purely reactive systems typically hard to debug
 - Lack of clear design methodology
 - (although learning control strategy is possible)
 - How about communication with humans?
- One final remark: don't confuse deliberative/reactive with symbolic/sub-symbolic (e.g. neural networks/genetic algorithms/numerical AI)

Hybrid Architectures

- Idea: Neither completely deliberateve nor completely reactive architectures are suitable - combine both perspectives in one architecture
- Most obvious approach: Construct an agent that exists of one (or more) reactive and one (or more) deliberative sub-components
- Reactive sub-components would be capable to respond to world changes without any complex reasoning and decision-making
- Deliberative sub-system would be responsible for abstract planning and decision-making using symbolic representations



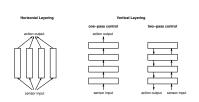
- Meta-level control of interactions between these components becomes a key issue in hybrid architectures
- Commonly used: lavered approaches
- Horizontal layering:
 - All layers are connected to sensory input/action output
 - Each layer produces an action, different suggestions have to be reconciled
- Vertical layering:
 - Only one layer connected to sensors/effectors
 - Filtering approach (one-pass control): propagate intermediate decisions from one laver to another
 - Abstraction layer approach (two-pass control): different layers make decisions at different levels of abstraction Informatics UoE Knowledge Engineering

Knowledge Engineering

Touring Machines

- Horizonal lavering architecture
- Three sub-systems: Perception sub-system, control sub-system and action sub-system
- Control sub-system consists of
 - Reactive laver: situation-action rules
 - Planning layer: construction of plans and action selection
 - Modelling layer: contains symbolic representations of mental state of other agents
- The three layers communicate via explicit control rules

Hybrid Architectures



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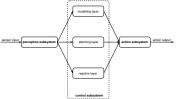
Touring Machines

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Reactive Architecture
Hybrid Architecture
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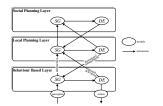
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Hybrid Architectures

InteRRaP



InteRRaP

- Vertical (two-pass) layering architecture
- InteRRaP: Integration of rational planning and reactive behaviour
- ► Three layers:
 - ▶ Behaviour-Based Layer: manages reactive behaviour of agent
 - Local Planning Layer: individual planning capabilities
 - Social Planning Layer: individual planning capabilities
 Social Planning Layer: determining interaction/cooperation
- strategies

 Two-pass control flow:
 - Upward activation: when capabilities of lower layer are exceeded, higher layer obtains control
 - Downward commitment: higher layer uses operation primitives of lower layer to achieve objectives

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Deliberation Architecture
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InteRRaP

- Every layer consists of two modules:
 - situation recognition and goal activation module (SG)
 - decision-making and execution module (DE)
- ► Every layer contains a specific kind of knowledge base

 ► World model
 - Mental model
 - Social model
- ► Only knowledge bases of lower layers can be utilised by any one layer
- Very powerful and expressive, but highly complex!

Summary

- Agent architectures: deliberative, reactive and hybrid
- ▶ Tension between reactivity and proactiveness
- ▶ BDI architecture: "intentional stance", computationally heavy
- ▶ Subsumption architecture: effective, but success sometimes "obscure"
- ▶ Hybrid architecture: attempt to balance both aspects, but increased complexity
- ▶ Next time: Agent interaction & communication

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