

Lecture 1

The Intelligent Agent Framework

Friday 22 August 2003

William H. Hsu

Department of Computing and Information Sciences, KSU

<http://www.kddresearch.org>

<http://www.cis.ksu.edu/~bhsu>

Reading for Next Class:
Chapter 2, Russell and Norvig

Lecture Outline

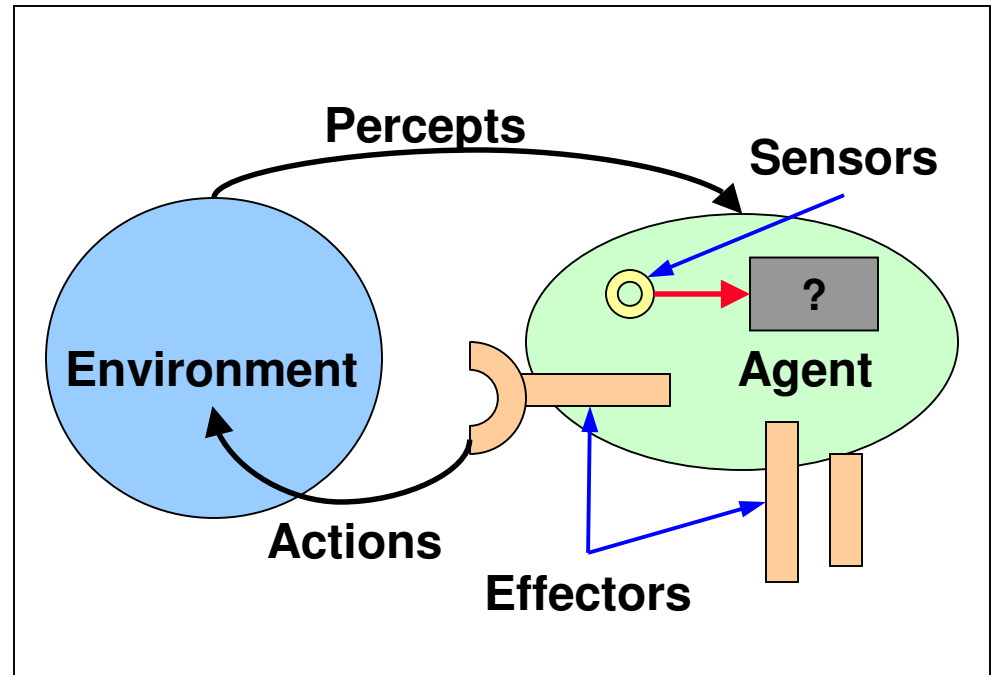
- **Today's Reading: Chapter 2, Russell and Norvig**
- **Intelligent Agent (IA) Design**
 - Shared requirements, characteristics of IAs
 - Methodologies
 - Software agents
 - Reactivity vs. state
 - Knowledge, inference, and uncertainty
- **Intelligent Agent Frameworks**
 - Reactive
 - With state
 - Goal-based
 - Utility-based
- **Thursday: Problem Solving and Search**
 - State space search handout (Winston)
 - Search handout (Ginsberg)

Review: Course Topics

- **Overview: Intelligent Systems and Applications**
- **Artificial Intelligence (AI) Software Development Topics**
 - Knowledge representation
 - Logical
 - Probabilistic
 - Search
 - Problem solving by (heuristic) state space search
 - Game tree search
 - Planning: classical, universal
 - Machine learning
 - Models (decision trees, version spaces, ANNs, genetic programming)
 - Applications: pattern recognition, planning, data mining and decision support
 - Topics in applied AI
 - Computer vision fundamentals
 - Natural language processing (NLP) and language learning survey
- **Implementation Practicum – 1-2 Students per Team**

Intelligent Agents: Overview

- **Agent: Definition**
 - Any entity that perceives its environment through sensors and acts upon that environment through effectors
 - Examples (class discussion): human, robotic, software agents
- **Perception**
 - Signal from environment
 - May exceed sensory capacity
- **Sensors**
 - Acquires percepts
 - Possible limitations
- **Action**
 - Attempts to affect environment
 - Usually exceeds effector capacity
- **Effectors**
 - Transmits actions
 - Possible limitations



How Agents Should Act

- **Rational Agent: Definition**

- Informal: “does the right thing, given what it believes from what it perceives”
- What is “the right thing”?
 - First approximation: *action that maximizes success of agent*
 - Limitations to this definition?
- Issues to be addressed now
 - How to evaluate *success*
 - When to evaluate *success*
- Issues to be addressed later in this course
 - Uncertainty (in environment, in actions)
 - How to express beliefs, knowledge

- **Why Study Rationality?**

- Recall: aspects of intelligent behavior (last lecture)
 - Engineering objectives: optimization, problem solving, decision support
 - Scientific objectives: modeling correct inference, learning, planning
- Rational cognition: formulating *plausible* beliefs, conclusions
- Rational action: “doing the right thing” given beliefs

Rational Agents

- **“Doing the Right Thing”**
 - Committing actions
 - Limited to set of effectors
 - In context of what agent knows
 - Specification (cf. software specification)
 - Preconditions, postconditions of operators
 - Caveat: not always perfectly known (for given environment)
 - Agent may also have limited knowledge of specification
- **Agent Capabilities: Requirements**
 - Choice: select actions (and carry them out)
 - Knowledge: represent knowledge about environment
 - Perception: capability to sense environment
 - Criterion: *performance measure to define degree of success*
- **Possible Additional Capabilities**
 - Memory (internal model of state of the world)
 - Knowledge about effectors, reasoning process (reflexive reasoning)

Measuring Performance

- **Performance Measure: How to Determine Degree of Success**
 - Definition: *criteria that determine how successful agent is*
 - Clearly, varies over
 - Agents
 - Environments
 - Possible measures?
 - Subjective (agent may not have capability to give accurate answer!)
 - Objective: *outside observation*
 - Example: web crawling agent
 - Number of hits
 - Number of *relevant* hits
 - *Ratio* of relevant hits to pages explored, resources expended
 - Caveat: “you get what you ask for” (issues: redundancy, etc.)
- **When to Evaluate Success**
 - Depends on objectives (short-term efficiency, consistency, etc.)
 - Is task episodic? Are there milestones? Reinforcements? (e.g., games)

Knowledge in Agents

- **Rationality versus Omniscience**

- Nota Bene (*NB*): not the same

- Distinction

- Omniscience: knowing *actual* outcome of all actions

- Rationality: knowing *plausible* outcome of all actions

- Example: is crossing the street to greet a friend too risky?

- Key question in AI

- *What is a plausible outcome?*

- Especially important in knowledge-based expert systems

- Of practical important in planning, machine learning

- Related questions

- *How can an agent make rational decisions given beliefs about outcomes of actions?*

- *Specifically, what does it mean (algorithmically) to “choose the best”?*

- **Limitations of Rationality**

- Based only on what agent *can* perceive and do

- Based on what is “likely” to be right, not what “turns out” to be right



What Is Rational?

- **Criteria**
 - Determines what is rational *at any given time*
 - Varies with agent, environment, *situation*
- **Performance Measure**
 - Specified by outside observer or evaluator
 - Applied (consistently) to (one or more) IAs in given environment
- **Percept Sequence**
 - Definition: *entire history* of percepts gathered by agent
 - NB: may or may not be retained fully by agent (issue: state and memory)
- **Agent Knowledge**
 - Of environment – “required”
 - Of self (reflexive reasoning)
- **Feasible Action**
 - What can be performed
 - What agent believes it can attempt?

Ideal Rationality

- **Ideal Rational Agent**
 - Given: any possible percept sequence
 - Do: ideal rational behavior
 - Whatever action is expected to maximize performance measure
 - NB: expectation – informal sense (for now); mathematical foundation soon
 - Basis for action
 - Evidence provided by percept sequence
 - Built-in knowledge possessed by the agent
- **Ideal Mapping from Percepts to Actions**
 - Figure 2.2, R&N
 - Mapping p : *percept sequence* \rightarrow *action*
 - Representing p as list of pairs: infinite (unless explicitly bounded)
 - Using p : specifies ideal mapping from percepts to actions (i.e., ideal agent)
 - Finding explicit p : in principle, could use trial and error
 - *Other (implicit) representations may be easier to acquire!*

Autonomy

- **Built-In Knowledge**
 - *What if agent ignores percepts?*
 - **Possibility**
 - All actions based on agent's own knowledge
 - Agent said to *lack autonomy*
 - **Examples**
 - “Preprogrammed” or “hardwired” industrial robots
 - Clocks
 - Other sensorless automata
 - NB: to be distinguished from closed versus open loop systems
- **Justification for Autonomous Agents**
 - Sound engineering practice: “Intelligence demands robustness, adaptivity”
 - Example: dung beetle (Egyptian scarab)
 - Ethological and evolutionary bases of knowledge
 - This course: mathematical and CS basis of autonomy in IAs

Structure of Intelligent Agents

- **Agent Behavior**
 - Given: sequence of percepts
 - Return: IA's actions
 - Simulator: description of results of actions
 - Real-world system: committed action
- **Agent Programs**
 - Functions that implement p
 - Assumed to run in computing environment (architecture)
 - Hardware architecture: computer organization
 - Software architecture: programming languages, operating systems
 - *Agent = architecture + program*
 - This course (CIS730): primarily concerned with p
 - CIS540, 740, 748: concerned with architecture
 - See also: Chapter 24 (Vision), 25 (Robotics), R&N
- **Discussion: “Real” versus “Artificial” Environments**

Agent Programs

- **Software Agents**
 - Also known as (aka) software robots, softbots
 - Typically exist in very detailed, unlimited domains
 - Example
 - (Real-time) critiquing, automation of avionics, shipboard damage control
 - Indexing (spider), information retrieval (IR; e.g., web crawlers) agents
 - Plan recognition systems (computer security, fraud detection monitors)
 - See: Bradshaw (*Software Agents*)
- **Focus of This Course: Building IAs**
 - Generic skeleton agent: Figure 2.4, R&N
 - function *SkeletonAgent* (*percept*) returns action
 - static: *memory*, agent's memory of the world
 - *memory* ← *Update-Memory* (*memory*, *percept*)
 - *action* ← *Choose-Best-Action* (*memory*)
 - *memory* ← *Update-Memory* (*memory*, *action*)
 - return *action*

Example: Automated Taxi Driver

- **Agent Type: Taxi Driver**
- **Percepts**
 - Visual: cameras
 - Profilometer: speedometer, tachometer, odometer
 - Other: GPS, sonar, interactive (microphone)
- **Actions**
 - Steer, accelerate, brake
 - Talk to passenger
- **Goals**
 - Safe, legal, fast, comfortable
 - Maximize profits
- **Environment**
 - Roads
 - Other traffic, pedestrians
 - Customers
- **Discussion: Performance Requirements for Open Ended Task**

Terminology

- **Artificial Intelligence (AI)**
 - **Operational definition**: study / development of systems capable of “thought processes” (reasoning, learning, problem solving)
 - **Constructive definition**: expressed in artifacts (design and implementation)
- **Intelligent Agents**
- **Topics and Methodologies**
 - **Knowledge representation**
 - Logical
 - Uncertain (probabilistic)
 - Other (rule-based, fuzzy, neural, genetic)
 - **Search**
 - **Machine learning**
 - **Planning**
- **Applications**
 - Problem solving, optimization, scheduling, design
 - Decision support, data mining
 - Natural language processing, conversational and information retrieval agents
 - Pattern recognition and robot vision

Summary Points

- **Artificial Intelligence: Conceptual Definitions and Dichotomies**
 - Human cognitive modelling vs. rational inference
 - Cognition (thought processes) versus behavior (performance)
 - Some viewpoints on defining intelligence
- **Roles of Knowledge Representation, Search, Learning, Inference in AI**
 - Necessity of KR, problem solving capabilities in intelligent agents
 - Ability to reason, learn
- **Applications and Automation Case Studies**
 - Search: game-playing systems, problem solvers
 - Planning, design, scheduling systems
 - Control and optimization systems
 - Machine learning: pattern recognition, data mining (business decision support)
- **More Resources Online**
 - Home page for AIMA (R&N) textbook
 - CMU AI repository
 - KSU KDD Lab (Hsu): <http://www.kddresearch.org>
 - Comp.ai newsgroup (now moderated)