What is an Agent?

Wide range of behavior and functionality in computing

Active computational entity

- With a persistent identity
- Able to carry out a long-lived conversation
- Perceives, reasons about, and initiates activities in its environment
 - Deals with services
- Communicates (with other agents)
 - Loosely coupled

Adaptive

Agents and Multiagent Systems for Services

Business partners are supported by agents

- Unlike objects, agents
 - Are proactive and autonomous—can say No!
 - Support loose coupling
- In addition, agents may
 - Cooperate or compete
 - Model users, themselves, and others
 - Dynamically use and reconcile ontologies

Modeling Agents: Artificial Intelligence

Emphasize mental (folk psychology) concepts to achieve simplicity of description

- Beliefs: agent's representation of the world
- Knowledge: (usually) true beliefs
- Desires: preferred states of the world
- ► Goals: consistent desires
- Intentions: goals adopted for action
 - Resources allocated
 - Sometimes incorporate persistence

Modeling Agents: Multiagent Systems

Emphasize interaction and autonomy and, hence, communication)

- Social: about collections of agents
- Organizational: about teams and groups
- Legal: about contracts and compliance
- Ethical: about right and wrong actions

Mapping Service-Oriented Computing to Agents

Agents capture the constraints of an open system

- ► Autonomy ⇒ ability to enter into and enact contracts
 - Counterbalanced by establishing compliance
 - How can we check or enforce compliance?
- Heterogeneity \Rightarrow ontologies
- ► Loose coupling ⇒ communication
- ► Trustworthiness ⇒ contracts, ethics, learning, incentives
- ► Dynamism ⇒ break and form relationships via combinations of the above

Two Main Ways to Apply Agents Agent-Oriented Software Engineering (AOSE)

As modeling constructs

- Standing in for stakeholders
- To help in capturing their requirements as goals
- As runtime constructs, each
 - Representing a stakeholder
 - Acting on its behalf, reflecting its autonomous decision making to others

Economic Rationality

Applies to business services

- Three elements: an agent's
 - Performance measure (for itself), e.g., expected utility
 - Prior knowledge and current (ongoing) perceptions
 - Available actions
- Ideally, for each possible percept sequence, a rational agent
 - Acts to maximize its expected utility
 - On the basis of its knowledge and evidence from the percept sequence

Logic-Based Agents

Logical reasoning being a form of rationality

An agent is a knowledge-based system

- Represents a symbolic (as opposed to neural) model of the world
- Declarative, hence, inspectable
- Reasons symbolically via logical deduction

Challenges:

- Representing information symbolically
 - Easier in information environments than in general
- Maintaining an adequate model of the world

Cognitive Architecture for an Agent

Sensors and effectors map to services
 Communication infrastructure is messaging middleware

Exercise

Create an instance of the preceding diagram where the two agents are Amazon and a manufacturer

- When is it beneficial to employ agents in this setting?
- What is an illustration of loose coupling in this setting?

A Reactive Agent

The Sense-Decide-Act Loop

```
Environment e;
RuleSet r; //Could be the receive method of an actor
while (true) {
  state = senseEnvironment(e);
  a = chooseAction(state, r);
  e.applyAction(a);
}
```

Generic BDI (Belief-Desire-Intention) Architecture

Addresses how beliefs, desires and intentions are represented, updated, acted upon Variant with just beliefs and goals is also prominent

```
Agent::run() {
Perception p;
p.run(); //start perception in own thread
while (true) {
  intention = getBestPlan();
  if (intention.execute()) // if achieved
    desires.remove(intention);
Perception :: run() {
while (true) {
  a. beliefs.incorporateNewObservations(getInput(w));
  if (! a.currentPlanlsApplicable())
    a.stopCurrentPlan();
  sleep(someShortTime);
```

Richer than sense-decide-act: decisions directly affect future decisions

Representing Services for Planning

IOPE (sometimes IOPR), goes beyond typical input-output signature

- Inputs: information the service requires
- Outputs: information the service produces
- Preconditions: constraints on the input
- Effects: effects on the environment
- Results (variant of effects): properties of the output

Composition as Planning

- Represent initial and goal states
- Represent each service as an action
 - Based on its IOPE specification
- A composed service: a plan that invokes constituent services
 - Inputs: outputs of previous services
 - Preconditions: true in initial state or made true by effects (results) of previous services
 - Effects not undone by subsequent services yield the goal state

Rules: Logical Representations

Marry declarative representation with computing

- Modular: easy to read and maintain
- Inspectable (by fact of being declarative): easy to understand
- Executable: no further translation needed
- Expressive: (commonly) Turing complete
 - Capture knowledge that would otherwise not be captured declaratively
 - Compare with relational calculus (classical SQL) or description logics (OWL)
- Declarative, although imperfectly so
 - Conflict handling is nontrivial and often ad hoc

Agents

Kinds of Rules

ECA (Event-Condition-Action) or Reaction

on event if condition then perform action

Derivation rules: special case of above, e.g., integrity constraints: derive false if error

Inference rules

 if antecedent
 then consequent

- Support multiple computational strategies
- Forward chaining; backward chaining

Agents

Architecture of an ECA-Based Agent



Applying ECA Rules

- Capture protocols, enterprise policies, and heuristics as ECA rules
 Examples?
- Combine with inference rules (to check if a condition holds)
- Modeling challenge
 - What is an event?
 - How to capture composite events by pushing event detection to lower layers

Example: ECA Rule

Identify predicates, variables, the do command, connectives

- Watch out for relevant events
- If one occurs, check condition
- If condition holds, perform action

Example: Inference Rule

Typical syntax indicating forward chaining

IF parent(?x ?y) AND parent (?y ?z) // Antecedent THEN grandparent (?x ?z) // Consequent

Typical syntax indicating backward chaining INFER grandparent (?x ?z) // Consequent FROM parent(?x ?y) // Antecedent AND parent (?y ?z)

Example: Communication

Combining backward chaining and ECA

```
IF incoming-message(?x ?y ?z)
AND policyA(?x ?y ?w)
AND policyB(?x ?z ?v)
THEN send message(?x ?v ?w)
AND assert internal-fact(?x ?v ?w)
```

The policy stands for any internal decision making, usually defined as INFER policyA(?x ?y ?w) FROM ... INFER policyB(?x ?z ?v)

```
FROM ...
```

Exercise: Communication

State the customer's rules to capture how it might interact with a merchant in a purchase protocol

- ► RFQ: request for quotes
- (Price) quote
- Accept or Reject
- Goods
- Payment
- Receipt

Typical Rule Syntax Limitations

- Antecedent may have conjunction or disjunction
- Antecedent may have generally not have negation (nontrivial)
- Consequent may have conjunction but not disjunction or negation—to avoid ambiguity of what to do
- Rules are not nested
- Generally no else clause

Applying Inference Rules

- Capture requirements naturally
- Elaboration tolerance requires defeasibility
 - Conclusions are not firm in the face of new information
 - Formulate general rules
 - Override rules to specialize them as needed
- Leads to logical nonmonotonicity
 - Easy enough operationally but difficult to characterize mathematically
 - Details get into logic programming with negation

Negation and Nonmonotonicity

Strong negation, indicating falsity (i.e., nontruth)

- Traditional, two-valued logic
- Law of the excluded middle
- Weak negation, indicating absence of knowledge or absence of proof (depending upon the setting)
 - Goes beyond traditional, two-valued logic
 - A proposition and its strong negation may both be unknown

Nonmonotonicity

- Conclusions are retracted in light of additional information
- Common in real-life reasoning
- Not supported by traditional logic
- Weak negation is an early approach to achieve nonmonotonicity

Agents

Conflicts and Priorities

- Rules can, and frequently do, conflict
 - An outcome of modular knowledge acquisition
 - Inadvertently enable two rules with contradictory conclusions
- Solution: expand rules to contain all applicable exception conditions
 - Unwieldy rules
 - Must redo each time new rules are stated
 - Can be impossible for users to understand ⇒ a major motivation for rules in the first place
- Solution: assert which rule overrides another rule
 - Specificity based on predicates used: only generic basis for prioritizing one rule over another
 - Doesn't always apply
 - Rely on order in the rules program
 - Such an order may not exist
 - Nontrivial to maintain
 - Assert numeric (or categorical) weights on rules
 - Nontrivial to maintain
 - Assert rankings between rules
 - Nontrivial to maintain

Variables in Rules

For safety, do not introduce variables in action or consequent

- ECA rules introduce variables in event and condition
 - Free variable in action indicates perform action for each binding
- Inference rules introduce variables in antecedent
 - Free variable in consequent means assert it for each binding

Agents Summary

- Agents match requirements of open environments
- Agents go beyond objects and procedural programming
- Agent abstractions help express requirements in a natural manner
- Cognitive constructs for agents can be powerful
- Rules provide a simple means to construct information agents