CS486/686: Introduction to Artificial Intelligence Lecture 2 - Agents and Abstraction

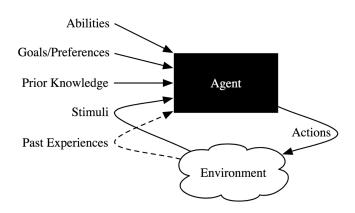
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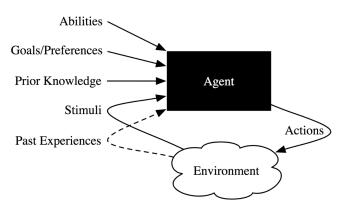
January 6, 2025

Readings: Poole & Mackworth 1.1

Situated Agent



Situated Agent



- An agent is an entity that performs actions in its environment
- Agent+Environment=world
- Inside black box: belief state

Four Example Application Domains (From Book)

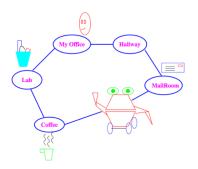
- Autonomous delivery robot roams around an office environment and delivers coffee, parcels,...
- Diagnostic assistant helps a human troubleshoot problems and suggests repairs or treatments, e.g., electrical problems and medical diagnosis
- Intelligent tutoring system teaches students in some subject area
- Trading agent buys goods and services on your behalf

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Let's talk about the autonomous delivery robot

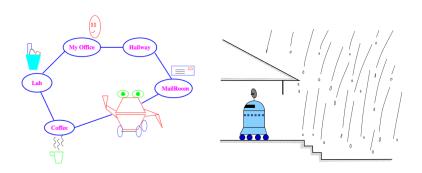
Domain for Delivery Robot



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Deliver coffee & mail when needed

Domain for Delivery Robot



The robot must:

- Deliver coffee & mail when needed
- Avoid getting wet

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- Goals: what it needs to deliver and when, tradeoffs between acting quickly and acting safely, effects of getting wet

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- Learn from experience
- Sense and act in the world, avoid obstacles, pickup and put down coffee, deliver mail

Knowledge Representation

- Knowledge: information used to solve tasks
- Representation: data structures used to encode knowledge
- Knowledge base (KB): representation of all knowledge
- Model: relationship of KB to world
- Level of Abstraction: How accurate is the model

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- Non-AI:
 - Specify how to compute something
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 - Programmer figures out how to do the computation
- AI:
 - Specify what needs to be computed
 - Specify how the world works
 - Agent figures out how to do the computation

Dimensions of Complexity

- The textbook defines 9 "dimensions of complexity" that make up the agent design space
- This is not a complete set (i.e., designing an agent is not as simple as selecting an option for each dimension)
- Can be multiple values in a dimension: values go from simple to complex
- Much of the history of AI can be seen as starting from the simple and adding in complexity in some of these dimensions

Dimensions of Complexity

- 1. **Modularity**: flat \rightarrow modular \rightarrow hierarchical
- Planning horizon: non-planning → finite horizon → indefinite horizon → infinite horizon
- Representation: explicit states → features → individuals and relations
- 4. Computational limits: perfect rationality \rightarrow bounded rationality
- 5. **Learning**: knowledge is given \rightarrow knowledge is learned
- Uncertainty: fully observable → partially observable world dynamics: deterministic → stochastic
- 7. **Preference**: goals \rightarrow complex preferences
- 8. Reasoning by number of agents: single agent o adversarial o multiagent
- 9. **Interactivity**: offline → online

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• Static: world does not change

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- Infinite horizon: the agent plans for going on forever (process oriented)

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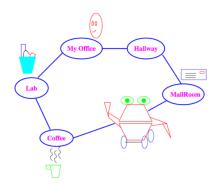
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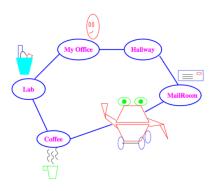
- Explicit states—a state directly represents one way the world could be
- Features or propositions
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 - 30 binary features can represent $2^{30} = 1,073,741,824$ states

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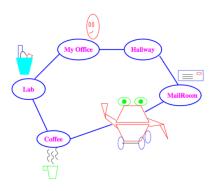
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- Individuals and relations
 - There is a feature for each relationship on each tuple of individuals
 - Often we can reason without knowing the individuals or when there are infinitely many individuals

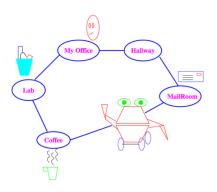




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- Features: robot location, user location, robot has coffee?, ...
- Relations: robot moves (clockwise + or counter-clockwise -) $\forall m \in \{+, -\}, \ell \in \{1, 2, 3...\}, move(m) : \ell' \leftarrow (\ell \ m \ 1)\%5$

Dimension 4: Computational Limits

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- Perfect rationality:
 Agent always chooses the optimal (i.e. highest utility) action
- Bounded rationality:
 Agent chooses a possibly sub-optimal action given its limited computational capacity (e.g. chess bots)
 - Satisficing solution (good enough)
 - Approximately optimal solution (how far off?)

Dimension 5: Learning from Experience

Whether the model is fully specified a priori:

- Knowledge is given
- Knowledge is learned from data or past experience

Dimension 6: Uncertainty

What the agent can determine the state from the observations:

- Fully-observable: the agent knows the state of the world from the observations
- Partially-observable: there can be many states that are possible given an observation

Uncertain World Dynamics

If the agent knew the initial state and the action, could it predict the resulting state?

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The dynamics can be:

- **Deterministic**: the state resulting from carrying out an action in state is determined from the action and the state
- **Stochastic**: there is uncertainty over the states resulting from executing a given action in a given state

Dimension 7: Goals or Complex Preferences

- Achievement goal is a goal to achieve; can be a complex logical formula
- Maintenance goal is a goal to be maintained
- Complex preferences that may involve tradeoffs between various desiderata, perhaps at different times; can be either ordinal or cardinal (e.g., utility)
- Examples: coffee delivery robot, medical doctor

Example: Complex Preferences

Delivery Robot



Example: Complex Preferences

Delivery Robot



- Goals may conflict
 - e.g. can't deliver mail and coffee at the same time
- Goals may be combinatorial
 - e.g. user may not want coffee if he doesn't get mail
- Goals may change
 - e.g. when wet, robot can't deliver mail
 - user switches from coffee to kale juice

Dimension 8: Reasoning by Number of Agents

- Single agent reasoning is where an agent assumes that any other agents are part of the environment (delivery robot)
- Adversarial reasoning considers another agent that acts in opposition to our goals (AlphaGo)
- Multiagent reasoning is when an agent needs to reason strategically about the reasoning of other agents (robot soccer, trading agents)

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Agents can have their own goals: cooperative, competitive, or goals can be independent of each other

Next

• Uninformed Search (Poole & Mackworth chapter 3)