CS344M Autonomous Multiagent Systems Spring 2008

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Good Afternoon, Colleagues



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Are there any questions?





• Reading responses



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 - Sorry about textbook availability!



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- Programming assignment



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 - Brooks' reactive robots
 - A more deliberative architecture
 - RoboCup case study



- Reading responses
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- Next week's readings are up:
 - Brooks' reactive robots
 - A more deliberative architecture
 - RoboCup case study
- Class discussions start Thursday
 Finalize after the first few



Franklin and Graesser Definition

An autonomous agent is a system situated within and a part of an environment that senses that environment and acts on it, over time, in pursuit of its own agenda and so as to affect what it senses in the future.



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1. "over time...so as to affect what it senses in the future"

2. "in pursuit of its own agenda"



- Are they agents or not?
- How does Wooldridge resolve this?



- They must **sense** their environment.
- They must **decide** what action to take ("think").
- They must **act** in their environment.



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Multiagent systems: Interact with other agents



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Learning agents: Improve performance from experience



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Multiagent systems: Interact with other agents

Learning agents: Improve performance from experience

Autonomous Bidding, Cognitive Systems, Traffic management, **Robot Soccer**



Your Agent Examples



Automotive: Anti-lock brake system (2); Engine Control Unit; Adaptive cruise control; Traffic light. **Physical Control:** Automatic sprinkler system; Auto coffee maker; Washing machine; Electric Guitar **Software Control:** File system indexer; Bittorent client; Auto song shuffler for radio Human monitoring: Pacemaker Game/entertainment: First Person Shooter enemy; Poker bot Service: Stock trading agent



Knowns:



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- $\mathcal{O} = \{\text{Blue}, \text{Red}, \text{Green}, \text{Black}, \ldots\}$
- Rewards in ${\sf I\!R}$
- $\mathcal{A} = \{Wave, Clap, Stand\}$

 $o_0, a_0, r_0, o_1, a_1, r_1, o_2, \ldots$



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Unknowns:

- S = 4x3 grid
- $\mathcal{R}: \mathcal{S} \times \mathcal{A} \mapsto \mathbb{R}$
- $\mathcal{P} = \mathcal{S} \mapsto \mathcal{O}$
- $\mathcal{T}: \mathcal{S} \times \mathcal{A} \mapsto \mathcal{S}$



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 $o_i = \mathcal{P}(s_i)$ $r_i = \mathcal{R}(s_i, a_i)$ $s_{i+1} = \mathcal{T}(s_i, a_i)$



Environments

Environment \implies sensations, actions



Environments

${\sf Environment} \Longrightarrow {\sf sensations}, {\sf actions}$

- fully observable vs. partially observable (accessible)
- deterministic vs. non-deterministic
- episodic vs. non-episodic
- static vs. dynamic
- discrete vs. continuous
- single-agent vs. multiagent



The Decision



• reactive vs. deliberative



- reactive vs. deliberative
- multiagent reasoning?



- reactive vs. deliberative
- multiagent reasoning?
- learning?



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• Standard agent:

 $action: \mathcal{S}^* \mapsto \mathcal{A}$



- $action : \mathcal{P} \mapsto \mathcal{A}$
- Decision based entirely on the present



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Reactive agents for Thursday's assignment task?



What new autonomous do you expect to see in the next 10 years?

