CS378 Autonomous Multiagent Systems Spring 2005

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Week 13a: Tuesday, April 18th

Good Afternoon, Colleagues

Are there any questions?





• Final tournament: Tuesday, May 16th, 1pm





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- All readings up





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- Final projects due in 2 weeks!



Recursive Modeling Method

• What should I do?



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- What should I do?
- What should I do given what I think you'll do?



Recursive Modeling Method

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- What should I do given what I think you'll do?
- What should I think you'll do given what I think you think I'll do?



Recursive Modeling Method

- What should I do?
- What should I do given what I think you'll do?
- What should I think you'll do given what I think you think I'll do?
- etc.



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 - Includes physical *and* mental states
 - Could be computationally expensive





Example: pursuit task

No-information: Random choice





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No-information: Random choice

Sub-intentional: Not rational

Intentional: Others use same model



Lessons

- Modeling can help
- There is a lot of useless information in recursive models
- Approximations (limited rationality) can be useful



• Use your own plans to model others



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- Use explicit team operators



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- Use your own plans to model others
- Use explicit team operators
 - Introduces challenges of role assignments, and
 - Minimum cost repair
- Assume agent is using a plan that you could use,
 - But not modeling you
- Act based on assumed actions of others



Where do Models Come From

Observation:

- Tambe and RMM: use existing model
 - No building a model



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What if we can't build a full model in advance?



Where do Models Come From

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What if we can't build a full model in advance?

• What are some incremental approaches for building a predictive model?



- Rock beats scissors
- Scissors beats paper
- Paper beats rock



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• What is your strategy before modeling me?



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- What is your strategy before modeling me?
- What is your strategy after modeling me?



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- What is your strategy before modeling me?
- What is your strategy after modeling me?
- Am I modeling you?



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- What is your strategy before modeling me?
- What is your strategy after modeling me?
- Am I modeling you?
- Would your end strategy change if I can?



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		Action	1	Action 2
Plaver 1	Action 1	1,0		3,2
	Action 2	2,1		4,0



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- Change the **best response** of the other agent



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Threats slides





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- Shoham:
 - 0-sum = single agent problem
 - common payoff = search for pareto optimum
 - General sum is the interesting case:
 - Learning in an environment with other, unknown, independent agents who may also be learning
 - Need to do well against some set of agents, never too poorly, and well against yourself.

