CS343 Artificial Intelligence

Prof: Peter Stone

Department of Computer Science The University of Texas at Austin

Good Morning, Colleagues



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Are there any questions?



- Late assignments accepted until the night before the exam (5/12)
- Final: Wednesday May 13th, 2pm-5pm
 - Open notes handwritten
 - No books, no printouts



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- Sample Dan Klein's Spring 2009 exam with solutions





Tournament Results

• 3. TeamTeam (Tres)



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- 2. CogentAgents (Colin)



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- 1. ArchimedesAndHerodotusAgent (Andrew and Paul)

Congratulations to all!



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- Al progress bound more by software or hardware?
- What if autonomous agents disagree?
- How would programs learn to learn?
- What does the scientific community think about the singularity?
 - Doing anything about it?
- Are AI researchers scared?



• Can machines be conscious?



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- Current progress are we "just" climbing a tree?
- Is AI is the right name for the field?

Would you have rather been born 100 years earlier or 100 years later?





• Does it matter to you if our "descendents" aren't human?



Peter Stone



• If an AI technology runs amok, who is responsible?



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- Are there some types of research we shouldn't do?



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- Will computers ever be better than people at everything?



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 - Also Bayesian networks for **classification**
 - A type of machine learning



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Maximize expected utility



Topics not covered

Constraint satisfaction

(Chapter 6)

- Knowledge representation and reasoning
 - (Chapters 7-9, 11, 12)

- Game theory and auctions
- Aspects of learning
- Natural language
- Vision

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Robotics



- (Chapters 18, 19)
- (Chapters 22, 23)
 - (Chapter 24)
 - (Chapter 25)



- Kim's and my surveys
- Positive and negative feedback useful



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- Invitation to send more feedback by email



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 - If/when I teach the course again, how should it change?



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- Most important: course rating, instructor rating, written comments



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