CS343
Artificial Intelligence

Prof: Peter Stone

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Good Morning, Colleagues

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Are there any questions?

Some Context

- **First weeks:** search (BFS, A*, minimax, alpha-beta)
  - Find an optimal plan (or solution)

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  - A type of **machine learning**

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  - **Week 11:** Neural nets and Deep Learning
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Topics not covered

- Knowledge representation and reasoning
  . (Chapters 7-9, 11, 12)
- Game theory and auctions (Sections 17.5, 17.6)
- Aspects of learning (Chapters 18, 19)
- Natural language (Chapters 22, 23)
- Vision (Chapter 24)
- Robotics (Chapter 25)