CS344M
Autonomous Multiagent Systems

Prof: Peter Stone

Department or Computer Science
The University of Texas at Austin

Good Afternoon, Colleagues

Are there any questions?

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- Final project proposal assigned

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Proposal (10/7): 3+ pages

- What you're going to do; graded on writing

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Due at beginning of classes

Overview of the Readings

Darwin: genetic programming approach

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Shafii: Open loop walk in 3d simulator

Evolutionary Computation

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Peter Stone

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  Some slides from *Machine Learning* (Mitchell, 1997)

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Architecture for Action Selection

- (other slides, video)

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- don't call "action selection"