# CS378 Autonomous Multiagent Systems Spring 2004

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Department of Computer Sciences The University of Texas at Austin

Week 6a: Tuesday, February 24th

### **Good Afternoon, Colleagues**

Are there any questions?



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• Soccer offsides





• Next week's readings posted





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- Final project have partners?



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



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#### Due at beginning of classes



Sahota: Reactive deliberation



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Darwin: 2 genetic programming approach



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**Darwin:** 2 genetic programming approach

**CMUnited99:** Teammate, opponent modeling; layered disclosure



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Riley: coaching



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- Riley: coaching
- Kok03: coordination graphs



- A hybrid approach
- Executor: carry out reactive behaviors
- Deliberator: evaluate possible high-level schema with parameters; generate bids
- Deliberator takes time, but something keeps happening always.
- In effect: deliberator commits to schema for some time



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



Arick Shao on evolving a soccer team

