CS395T Reinforcement Learning: Theory and Practice Fall 2004

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Week11b: Thursday, November 18th

Good Afternoon Colleagues

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- Pending questions:
 - Helicopter: MDP or POMDP?
 - Bagnell Theorem 2 correct?

Logistics

Tom Dietterich visiting tomorrow:
 "Three Challenges for Machine Learning Reserch"
 3pm, ACES 2.302

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Student-led Discussion

Michael on biological primitives

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