

**CS395T**  
**Agent-Based Electronic Commerce**  
**Fall 2006**

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Week 4a

# Good Afternoon, Colleagues

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- How can you measure progress
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- Realism: how well do designs transfer?
- Collusion: does it happen?
- Did agents know identities of others?
- Open vs. closed loop strategies
- Does reasoning about timing help?

# 28 Simultaneous Auction

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**Flights:** Inflight days 1-4, Outflight days 2-5 (8)

- Unlimited supply; prices random walk; immediate clear; no resale

**Hotels:** Tampa Towers/Shanties 1-4 (8)

- 16 rooms per auction; 16th-price ascending auction; quote is ask price; no resale
- Auctions can close early; “beat the quote”

**Entertainment:** MU/AP/AW days 1-4 (12)

- Continuous double auction; initial endowments; quote is bid-ask spread; resale allowed

# Client Preferences and Utility

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**Preferences:** randomly generated per client

- Ideal arrival, departure days
- Good Hotel Value
- Entertainment Values

**Utility:** 1000 (if valid) – travel penalty + hotel bonus  
+ entertainment bonus

**Score:** Sum of client utilities – expenditures



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- Three measures found to be significant:
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  3. ratio of “easy” days (1 and 4) to hard (2 and 3) in preferred trip intervals

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- Three measures found to be significant:
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  3. ratio of “easy” days (1 and 4) to hard (2 and 3) in preferred trip intervals
- Regression analysis to compute factors for individual games

# Acquisition

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Optimal complete itinerary assignments

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“Branch and bound” over adjustments for 3
- Globally optimal solution; *usually*  $< .01$  sec

# TAC 2000

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- ATTac and Roxybot

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- Did agents know identities of others?

# Controlled experiments from ATTac-2000

ATTac vs. non-adaptive high and low bidders

#high	agent 2	agent 3	agent 4	agent 5	agent 6	agent 7	agent 8
7	←	9526	→				
6	←	10679	→				1389
5	←	10310	→			←	2650
4	←	10005	→		←	4015	
3	←	5067	→		←	3639	
2	←	209		←	2710		

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- Shows ATTac's average score difference
- ATTac adapts over successive runs
- All numbers positive, most are significant



# Class Discussion

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- Todd on hotel price prediction

# AverageMU

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- ATTac01's strategy

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- Open vs. closed loop experiments

# SAA

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- Solve an approximation of the problem that incorporates only the sample scenarios.
- Heuristics defined in chapter 5 (book.pdf in same place)



# Axelrod's tournament

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- What if you play infinitely?
- What if you play for a known finite amount of time?
- Some strategies:
  - hawk (always Fink)
  - Grim trigger (cooperate until the other defects)
  - tit-for-tat
  - Joss (tit-for-tat with periodic defection)