# Game-theoretic Foundations of Multi-agent Systems

Lecture 1: Introduction

Seyed Majid Zahedi



#### Outline

1. Course Mechanics

2. Course Outline

3. Overview of Game Theory and Mechanism Design



- Course website
  - https://ece.uwaterloo.ca/~smzahedi/crs/ece750/
  - All course information, lecture notes, assignments, etc.





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- Anti-requisites
  - ECON.412 (Topics in Game Theory), CO.456 (Introduction to Game Theory), CS.886 (Multi-agent Systems), and MSCI.724 (Game Theory and Recent App)

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- Games in extensive form
  - Perfect and imperfect-information games, finite and infinite-horizon games, subgame-perfect equilibrium, backward induction, one-shot deviation principle

# Tentative Topics (cont.)

- Beyond normal and extensive-form games
  - Repeated games, stochastic games, Bayesian games, congestion games, trigger strategies, folk theorems, Bayes-Nash equilibrium, auctions

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- Learning in multi-agent systems
  - Multi-agent reinforcement learning, fictitious play, Bayesian learning, regret-minimization learning

#### Textbook and References

- Y. Shoham and K. Leyton-Brown, Multi-agent Systems: Algorithmic, Game-theoretic, and Logical Foundations (available online)
- N. Nisan, et al. Algorithmic Game Theory (available online)
- T. Roughgarden, Twenty Lectures on Algorithmic Game Theory (notes available online)
- D. Fudenberg and D. Levine, The Theory of Learning in Games
- D. Fudenberg and J. Tirole, Game Theory
- M. J. Osborne and A. Rubinstein, A Course in Game Theory (available online)



Graduate offering



- Graduate offering
  - Quizzes 5%



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# Survey Project (Undergraduate Offering)

- Goal is to review existing literature in sub-area of multi-agent systems and possibly explore open research questions in that sub-area
- Two milestones: proposal (15% about three weeks into the term) and final written report (85% end of the term)

# Research Project (Graduate Offering)

- Goal is to try to do something novel, rather than merely surveying existing work
- Only real constraint is that it has to have something to do with covered material
- Projects may be theoretical, experimental (based on simulations), experimental (based on real-world data), a useful software artifact, or any combination thereof
- Three milestones: proposal (10%), oral progress report (10%), and final written report (80%)
- Creativity is encouraged

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- Is there a role for (machine) learning in games?
- What types of modeling problems do we face when addressing real-world games?

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- What are the possible outcomes?
- Steady-state, stable operating point, characteristics?
- How do you get there (learning dynamics, computation of equilibrium)?

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- Everyone writes down an integer between 1 and 100
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- Example:
  - A says 50 B says 10 C says 90
  - 2/3 of average(50, 10, 90) =  $2/3 \times 50 = 33.33$
  - A is the closest (|50 33.33| = 16.67), so A wins

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- In Economics, MD is all about designing the right incentives
  - Mechanisms map signals from independent agents into allocations and payments
  - Optimal Mechanisms (Myerson): Design a mechanism that maximizes profits
  - Efficient Mechanisms (Vickrey-Clarke-Groves (VCG) Mechanisms): Design a mechanism to maximize a social or system-wide objective



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- In CS/Engineering, focus is more on the design of efficient decentralized protocols that take into account incentives
  - Mechanisms in networks, distributed and online mechanisms, mechanisms that operate with limited information

## Example: Single-item Auction

- Sealed-bid auction: every bidder submits bid in sealed envelope
- First-price sealed-bid auction: highest bid wins, pays amount of own bid
- Second-price sealed-bid auction: highest bid wins, pays amount of 2nd-highest bid

#### Which Auction Generates More Revenue?

- Each bid depends on
  - bidder's true valuation for the item (utility = valuation payment)
  - bidder's beliefs over what others will bid
  - and ... auction mechanism used
- In first-price auction, it does not make sense to bid your true valuation
  - Even if you win, your utility will be 0
- In second-price auction, it always makes sense to bid your true valuation (we will see this later)



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- Can we solve for the optimal mechanism?

## Acknowledgment

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  - Vincent Conitzer [Duke CPS 590.4]