



## MSML610: Advanced Machine Learning

### Game Theory

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**References:**

# Game theory

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- **Game theory**
  - Connections between Machine Learning and Game Theory

# Normal Form Games

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- Representation of games using a payoff matrix
- Players choose strategies simultaneously
- Used to study strategic interactions
- Example: Prisoner's Dilemma

# Extensive Form Games

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- Game represented as a tree with sequential moves
- Captures order of play, decisions, and information
- Includes concepts like subgames and information sets
- Used in modeling turn-based interactions (e.g., poker)

# Pure and Mixed Strategies

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- Pure strategy: deterministic choice
- Mixed strategy: probability distribution over actions
- Mixed strategies allow randomization and increase flexibility
- Mixed strategies are crucial in achieving equilibrium in some games

# Nash Equilibrium

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- A set of strategies where no player has an incentive to deviate
- Can be in pure or mixed strategies
- Existence guaranteed (in mixed strategies) by Nash's theorem
- Central concept in game theory

# Dominant and Dominated Strategies

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- A strategy dominates another if it always yields better payoff
- Dominated strategies are never rational to play
- Iterated elimination of dominated strategies simplifies games

# Best Response and Reaction Functions

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- Best response: strategy that maximizes a player's payoff given others' strategies
- Nash equilibrium occurs where all players play best responses
- Used to derive and analyze equilibria



# Zero-Sum Games

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- One player's gain is another player's loss
- Payoffs satisfy  $u_1 = -u_2$
- Solvable using minimax theorem
- Important in adversarial ML (e.g., GANs)

# Minimax Theorem

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- In zero-sum games:  $\min_{\pi_1} \max_{\pi_2} u = \max_{\pi_2} \min_{\pi_1} u$
- Players choose strategies to minimize their maximum possible loss
- Foundation for robust optimization and adversarial training

# Correlated Equilibrium

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- Players coordinate using a shared signal
- Generalization of Nash equilibrium
- Can lead to better outcomes than Nash equilibrium
- Enables learning equilibria via no-regret algorithms

# Bayesian Games

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- Games with incomplete information
- Players have private types and beliefs
- Strategies depend on beliefs about others
- Relevant to modeling uncertainty in ML

# Evolutionary Game Theory

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- Studies dynamics of strategy changes in populations
- Focuses on replicator dynamics and evolutionarily stable strategies (ESS)
- Used in multi-agent learning and biology-inspired algorithms

# Repeated Games

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- Same game played multiple times
- History of play influences future actions
- Cooperation can emerge (e.g., tit-for-tat)
- Important for modeling long-term interaction

# Stochastic Games

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- Generalization of repeated games with state transitions
- Combines game theory with Markov decision processes (MDPs)
- Used in reinforcement learning and dynamic competition

# Mechanism Design

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- Inverse game theory: design rules to achieve desired outcomes
- Ensures incentive compatibility and individual rationality
- Applied in auctions, resource allocation, and blockchain systems



# Auction Theory

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- Study of bidding strategies and auction formats
- Types: first-price, second-price (Vickrey), all-pay, etc.
- Truthful bidding is optimal in second-price auctions
- Used in online ad auctions and marketplaces

# Potential Games

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- Games with a potential function aligning with player incentives
- Any change in individual utility reflects a global change
- Facilitates convergence to equilibrium via learning

# Learning in Games

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- Fictitious play, regret minimization, Q-learning in multi-agent settings
- Focus on convergence to equilibrium through repeated interaction
- No-regret algorithms converge to correlated equilibria

# Multi-Agent Reinforcement Learning (MARL)

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- Agents learn by interacting with other learning agents
- Combines RL with game-theoretic reasoning
- Challenges: non-stationarity, exploration, coordination

# Cooperative Game Theory

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- Focuses on group outcomes and coalition formation
- Concepts: core, Shapley value, bargaining
- Useful in team-based ML and federated learning

# Social Choice and Voting

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- Aggregation of individual preferences into a collective decision
- Arrow's impossibility theorem: no perfect voting system exists
- Relevant in ensemble methods and preference aggregation

# Connections between Machine Learning and Game Theory

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- Game theory
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