Advanced Computer Networks SS 2016

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Outline of Wireless Block

- Game theory and its applications
- Game theory basics and concepts
- Distributed Spectrum Sharing Application
- Social Group Maximization Framework
- Introduction to the framework
- Wireless Network Applications
- Mobile Data Offloading
- Basics and ideas
- Optimized Offloading Decision

Final Exam in this block only covers basic concepts and examples

Introduction to Game Theory

Game Theory



Rational – user aims to optimize its own objective

Interaction – user needs to take others' decisions into account

- "...Game Theory is designed to address situations in which the outcome of a person's decision depends not just on how they choose among several options, but also on the choices made by the people they are interacting with..."-David Easley and Jon Kleinberg
- "... Game theory is the study of the ways in which <u>strategic</u> <u>interactions</u> among <u>rational</u> agents produce <u>outcomes</u> with respect to the <u>utilities</u> of those agents" --Stanford Encyclopedia of Philosophy

A Brief History

- 1944: Von Neumann and Oskar Morgenstern Theory of Games and Economic Behavior Two-player games
- 1950: John Nash

Nash Equilibrium Equilibrium points in n-player games

• After 1950s: widely used in economics, politics, biology...

Competition between firms

Auction design

Role of punishment in law enforcement

- International policies
- Evolution of species



O. Morgenstern 1902-1977



von Neumann 1903-1957



John Nash 1928-2015

Relevance to Networking Research

- Economic issues become increasingly important
 - Interactions with/between human users
 e.g., data plan pricing, resource allocation
 - Independent service providers
 e.g., bandwidth trading, peering agreements
- Tool for system design
 - Distributed algorithms
 - Multi-objective optimization
 - Incentive compatible protocols



A COURSE IN

MARTIN J. OSBORNE ARIEL RUBINSTEIN



DREW FUDENBERG AND JEAN TIROLI

Game Theory Basics

- Strategic game form (*P*, *S*, *U*)
- Players (P_1, \dots, P_N) : finite number of decision makers
- Strategy sets $(S_1, ..., S_N)$: player P_i has a nonempty set S_i of actions/strategies s_i
- Payoff function $U_i(s_1, ..., s_N)$: player's preference/individual utility
- Pure Nash equilibrium (NE)
- A strategy profile $(s_1^*, \dots, s_i^*, \dots, s_N^*)$ is a NE if for each player i $U_i(s_1^*, \dots, s_i^*, \dots, s_N^*) \ge U_i(s_1^*, \dots, s_i, \dots, s_N^*), \forall s_i \in S_i$
- No player has incentive to deviate (stable system point)
- NE is a fixed point of the best response functions

$$s_i^* = \underset{s_i \in S_i}{\operatorname{argmax}} U_i(s_1^*, \dots, s_i, \dots, s_N^*), \forall i$$

• There is no universal rule for finding a Nash equilibrium!

- Two suspects are arrested
- The police lack sufficient evidence to convict the suspects, unless at least one confesses
- The police hold the suspects in two separate rooms, and tell each of them three possible consequences:
 - If both deny: 1 month in jail each
 - If both confess: 6 months in jail each
 - If one confesses and one denies:
 - The one confesses: walk away free of charge
 - > The one denies: serve 12 months in jail



- Strictly dominated strategy
- Player *i*'s strategy s'_i is strictly dominated by player *i*'s strategy s_i if $U_i(s_i, s_{-i}) > U_i(s'_i, s_{-i}), \forall s_{-i}$

where s_{-i} is the strategy profile of all the other players except player *i*

- No matter what other people do, by choosing s_i instead of s'_i , player i will always obtain a better payoff
- Key principle: Never play a strictly dominated strategy



Deny is strictly dominated by Confess!

Finding Nash Equilibrium

- When there are no strictly dominated strategies, we can not easily "simplify" the game
- Nash equilibrium is a state of mutual best responses
- Key principle: derive the best responses

- Two hunters decide what to hunt independently
- Each one can hunt a stag (deer) or a hare
- Successful hunt of stag requires cooperation
- Successful hunt of hare can be done individually
- Simultaneous decisions without prior communications



There is no strictly dominated strategy Find out a player's best response given the other player's choice







- Two Nash equilibria exist
- (Stag, Stag) is payoff dominant
 - > Both players get the best payoff possible
 - Require trust among players to achieve coordination
- (Hare, Hare) is risk dominant
 - Minimum risk if player is uncertain of each other's choice

- A couple decide where to go during Friday night without communications
- Husband prefers to go and watch football
- Wife prefers to go and watch ballet
- Both prefer to stay together during the night



There is no strictly dominated strategy Find out a player's best response given the other player's choice







NE is a state of mutual best responses

- Prisoner's dilemma, Stag Hunt, Battle of Sexes are finite games with finite number of actions for each player
- Cournot competition is a continuous game in which a player has continuous (infinite) choices

- Two firms producing the same kind of product in quantities of q₁ and q₂, respectively
- Market clearing price $p = A q_1 q_2$
- Cost of unit production is *C* for both firms
- Profit for firm *i*

 $J_i = (p - C)q_i$ = $(A - C - q_1 - q_2)q_i$ Define $B \equiv A - C$

• Objective of firm *i*: choose q_i to maximize profit $q_i^* = \underset{q_i}{\operatorname{argmax}} (B - q_1 - q_2)q_i$

- Firm *i*'s best response, given its competitor's q_j $q_i^* = (B - q_j)/2$
- NE (q_1^*, q_2^*) of Cournot competition satisfies $\begin{cases}
 q_1^* = (B - q_2^*)/2 \\
 q_2^* = (B - q_1^*)/2
 \end{cases}$
- This leads to the NE as

 $q_1^* = q_2^* = B/3$

- Firm *i*'s best response, given its competitor's q_j $q_i^* = (B - q_j)/2$
- NE (q_1^*, q_2^*) of Cournot competition satisfies $\begin{cases}
 q_1^* = (B - q_2^*)/2 \\
 q_2^* = (B - q_1^*)/2
 \end{cases}$



Summary

- Game theory: players, strategies, and payoffs
- Nash equilibrium
- Strictly dominated strategy
- Best response strategy
- Finite and infinite(continuous) games

Application in Distributed Spectrum Sharing

- Spectrum is scarce
 - ► Most spectrums have been exclusively licensed
 - ► More and more wireless devices emerge



- Spectrum is under-utilized
 - ▶ E.g., average spectrum utilization in Chicago is lower than 20%

- Dynamic spectrum access with cognitive radios
 - Address spectrum under-utilization problem
 - ► Primary user (PU) licensed spectrum holder
 - ► Secondary user (SU) unlicensed spectrum user
 - ► Enable SUs opportunistically share the spectrum with PUs

Spectrum in use
Dynamic Spectrum Access



- Achieving efficient distributed spectrum sharing among SUs is challenging
 - ► Spectrum opportunities change over frequency, time, and space
 - Individual SU has limited observations of the entire network
 - Multiple simultaneous SUs may generate severe interference
- SUs are hence required to make intelligent decisions for efficient spectrum sharing

Key problem: how to share spectrum in an intelligent way?

- Individual intelligence
 - SUs share the spectrum competitively based on strategic interactions
 - ► SUs are fully rational (e.g., belong to different authorities)
 - ► Non-cooperative spectrum access games



Distributed Spectrum Sharing with Spatial Reuse

- SUs with individual intelligence share the spectrum based on competition
- Game theory for modeling competitive spectrum sharing
- Spatial aspect of competitive spectrum sharing is less understood
 Spatial reuse is a key feature of wireless communication
 - ► Most existing works focus on fully interfering case
- Spatial spectrum access game for competitive spectrum sharing with spatial reuse

System Model

- *M* heterogeneous channels
 - ▶ Channel state $S_m(\tau) \in \{0, 1\}$ with an idle probability of θ_m
 - ▶ $\mathcal{M} = \{1, 2, ..., M\}$ the set of channels
- *N* heterogeneous SUs
 - ▶ Data rate $b_m^n(\tau)$ follows i.i.d random process with mean B_m^n
 - ▶ $\mathcal{N} = \{1, 2..., N\}$ the set of SUs

System Model

- Interference graph $G = \{\mathcal{N}, \mathcal{E}\}$
 - \blacktriangleright \mathcal{N} vertex set
 - ► $\mathbf{d}_n = (d_{T_{X_n}}, d_{R_{X_n}})$ locations of transmitter and receiver of SU *n*
 - ► $\mathcal{E} = \{(i,j) : ||d_{T_{X_i}}, d_{R_{X_j}}|| \le \delta_i, \forall i, j \ne i \in \mathcal{N}\}$ interference edge set
 - ▶ $\mathcal{N}_n = \{i : (i, n) \in \mathcal{E}, i \in \mathcal{N}\}$ interfering SUs of SU n



System Model

- Channel contention model
 - ▶ $\mathbf{a} = (a_1, ..., a_N)$ channel selections of all SUs
 - ▶ $\mathcal{N}_n^{a_n}(\mathbf{a}) = \{i \in \mathcal{N}_n : a_i = a_n\}$ interfering SUs choosing the same channel as SU *n*
 - ▶ $g_n(\mathcal{N}_n^{a_n}(\mathbf{a}))$ probability that SU *n* can grab the channel
- Expected SU throughput

$$U_n(\mathbf{a}) = \theta_{a_n} B_{a_n}^n g_n(\mathcal{N}_n^{a_n}(\mathbf{a}))$$

Spatial Spectrum Access Game (SSAG)

• Each SU with individual intelligence tries to maximize its own throughput

 $\max_{a_n \in \mathcal{M}} U_n(a_n, a_{-n}), \forall n \in \mathcal{N}$

- Spatial spectrum access game (SSAG) $\Gamma = (\mathcal{N}, \mathcal{M}, G, \{U_n\}_{n \in \mathcal{N}})$
- Key problems:
 - Does SSAG have a Nash equilibrium?
 - ▶ If SSAG has an equilibrium, how to achieve it?

Spatial Spectrum Access Game (SSAG)

• Key problems:

► Does SSAG have a Nash equilibrium?

► If SSAG has an equilibrium, how to achieve it?

Existence of Nash Equilibrium

• Pure Nash equilibrium $\mathbf{a}^* = (a_1^*, ..., a_N^*)$

$$a_n^* = rg\max_{a_n} U_n(a_n, a_{-n}^*), \forall n \in \mathcal{N}$$

- ► SUs are mutually satisfactory (no SU can improve unilaterally)
- Mixed strategy σ_n = (σ₁ⁿ, ..., σ_Mⁿ)
 ⊳ σ_mⁿ probability of SU n choosing channel m
 ⊳ U_n(σ₁, ..., σ_N) SU throughput under mixed strategies
- Mixed Nash equilibrium $(\sigma_1^*, ..., \sigma_N^*)$

$$\boldsymbol{\sigma}_n^* = rg\max_{\boldsymbol{\sigma}_n} U_n(\boldsymbol{\sigma}_n, \boldsymbol{\sigma}_{-n}^*), \forall n \in \mathcal{N}$$

- Example: 2 SUs, 2 Channels, and $U_n(\mathbf{a}) = p \prod_{i \in \mathcal{N}_n^{a_n}(\mathbf{a})} (1-p)$
 - ▶ Pure Nash equilibrium: (1,2) and (2,1)
 - Mixed Nash equilibrium: $\left(\left(\frac{1}{2}, \frac{1}{2}\right), \left(\frac{1}{2}, \frac{1}{2}\right)\right)$

Existence of Nash Equilibrium

- SSAG always has a mixed Nash equilibrium
 - SSAG is a finite game
 - ► Finite game always has a mixed Nash equilibrium (John Nash 1950)
 - Require channel switching
- Pure Nash equilibrium requires no channel switching
- SSAG may not admit a pure Nash equilibrium
 - ► Example: 3 SUs, 2 channels, $U_n(\mathbf{a}) = p \prod_{i \in \mathcal{N}_n^{a_n}(\mathbf{a})} (1-p)$



• Graphical structure plays a key role

Lemma one:

- Suppose that SSAG on a directed graph G has a pure Nash equilibrium
- Construct a new SSAG by adding a new SU
 - ► Can not generate interference to SUs in original SSAG
 - ► May receive interference from SUs in original SSAG
- New SSAG game also has a pure Nash equilibrium



Theorem

SSAG on a directed acyclic graph has a pure Nash equilibrium.

- Directed acyclic graph can be given a topological sort (ordering of nodes)
- ► No edges directed from nodes of high order to nodes of low order



- Congestion property (CP): *N˜_n^{a_n}(a) ⊆ N^{a_n}_n(a) ⇒ g_n(<i>N˜_n^{a_n}(a)) ≥ g_n(N^{a_n}_n(a))* More contending SUs, less chance to grab the channel
 - ► Natural for practical wireless systems

Lemma two:

- Suppose that SSAG satisfying CP on a directed graph G has a pure Nash equilibrium
- Construct a new SSAG by adding a new SU
 - Channel contention satisfies CP
 - ► May generate/receive interference to/from at most one SUs in original SSAG
- New SSAG game also has a pure Nash equilibrium

Lemma two:

- Suppose that SSAG satisfying CP on a directed graph G has a pure Nash equilibrium
- Construct a new SSAG by adding a new SU
 - ► Channel contention satisfies CP
 - ► May generate/receive interference to/from at most one SUs in original SSAG
- New SSAG game also has a pure Nash equilibrium



Theorem

SSAG satisfying CP on a directed tree/forest has a pure Nash equilibrium.

Directed tree – the corresponding graph without directions is a tree



► Directed forest – a disjoint union of directed trees



• More interference graphs can be constructed from Lemmas one & two



Existence of Nash Equilibrium



Spatial Spectrum Access Game (SSAG)

• Key problems:

- ► Does SSAG have a Nash equilibrium?
- ► If SSAG has an equilibrium, how to achieve it?

Achieving Equilibrium For SSAG

- SSAG always has a mixed Nash equilibrium
- Sharing information is not incentive compatible
 SUs with individual intelligence are competitive
- How to achieve a mixed Nash equilibrium without information exchange?
- Distributed learning mechanism
 - Extend single-agent reinforcement learning to a multi-agent setting
 - Each SU estimates its throughput locally
 - Each SU adapts channel access strategy based on accumulated estimations

Distributed Learning For SSAG

- SU *n* chooses a channel $a_n(t)$ according to mixed strategy $\sigma_n(t) = (\sigma_n^1(t), ..., \sigma_n^M(t))$
- Mixed strategy $\sigma_n(t)$ is generated according to perceptions $\mathbf{P}_n(t) = (P_n^1(t), ..., P_n^M(t))$ as

$$\sigma_n^m(t) = \frac{\exp(\gamma P_n^m(t))}{\sum_{i=1}^M \exp(\gamma P_n^i(t))}$$

- *P*^m_n(t) − estimated performance of choosing channel m
 γ − balance between exploration and exploitation
- Perceptions $\mathbf{P}_n(t)$ are updated as

$$P_n^m(t+1) = \begin{cases} (1-\mu_t)P_n^m(t) + \mu_t \tilde{U}_n(\mathbf{a}(t)), & \text{if } a_n(t) = m \\ P_n^m(t), & \text{otherwise} \end{cases}$$

- ▶ μ_t smoothing factor
- Reinforce the perception of the channel just accessed

Convergence of Distributed Learning

• $\sigma = (\sigma_1, ..., \sigma_N)$ is an ϵ -approximate mixed Nash equilibrium if

$$U_n(\sigma_n, \sigma_{-n}) \geq \max_{\hat{\sigma}_n} U_n(\hat{\sigma}_n, \sigma_{-n}) - \epsilon, \forall n \in \mathcal{M}$$

▶ ϵ – gap from an exact mixed Nash equilibrium

Convergence of Distributed Learning

Convergence by stochastic approximation theory

• γ satisfies

$$\gamma < \frac{1}{2 \max_{m \in \mathcal{M}, n \in \mathcal{N}} \{\theta_m B_m^n\} \max_{n \in \mathcal{N}} \{|\mathcal{N}_n|\}}$$

▶ Smaller $\gamma \Longrightarrow$ more random exploration \Longrightarrow better environment understanding

•
$$\sum_t \mu_t = \infty$$
 and $\sum_t \mu_t^2 < \infty$

Theorem

Distributed learning for SSAG converges to an ϵ -approximate mixed Nash equilibrium σ^* with $\epsilon = \max_{n \in \mathcal{N}} \{-\frac{1}{\gamma} \sum_{m=1}^{M} \sigma_m^{n*} \ln \sigma_m^{n*} \}.$

ϵ ≤ 1/γ ln M, e.g., random channel access
 Convergence depends on the structure of interference graph
 Trade-off between exploration and exploitation: smaller γ for convergence ⇒larger performance gab ϵ

Simulation

• M = 5 channels and N = 9 SUs on four different interference graphs



Simulation

• Up-to 100% performance improvement over random access



- At most 28% performance loss over centralized optimal solution on Graph (b) without spatial reuse
- At most 10% performance loss over centralized optimal solution on Graphs (a), (c), and (d) with spatial reuse

Summary of Spatial Spectrum Access Game

- SSAG for competitive spectrum sharing with spatial reuse
- Explore the existence of both pure and mixed Nash equilibrium
- Distributed learning for achieving approximate mixed Nash equilibrium