

IE 598ns

Game theory: Models, Algorithms and Applications

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Introduction

- Course: Game theory: Models, Algorithms and Applications
- Focus on a variety of game-theoretic problem classes
- Restricted to problems with continuous strategy sets
- An optimization bias - cast game-theoretic problems as complementarity or variational inequality problems (or others)
- Useful viewpoint - provides us with a means to discuss existence and uniqueness of equilibria as well as tools to obtain such points

Example: Prisoner's dilemma

- Two suspects are arrested and charged with a crime
- Explain consequences:
 - if neither confesses: both will get 1 month in jail
 - if both confess: both will get 6 months in jail
 - if one confesses, he will be released while the other will get 9 months
- players simultaneously choose actions
- receive payoffs based on decisions of all players

- Bimatrix game:

		Prisoner 2	
		Mum	Fink
Prisoner 1	Mum	(-1,-1)	(-9,0)
	Fink	(0,-9)	(-6,-6)

Normal form representation

- The players in the game, indexed by $i \in \{1, \dots, N\}$
- Player i 's strategy $s_i \in S_i$, the strategy set of player i
- Payoff of player i is given by $u_i(s_1, \dots, s_N)$.

Definition 1 *The normal form representation of an N -player game specifying strategy spaces S_1, \dots, S_N and their payoffs u_1, \dots, u_N . We denote this game by $G = \{S_1, \dots, S_N; u_1, \dots, u_n\}$.*

- Game theory provides a unique prediction for each player's strategy
- Predicted strategy should be *best response* to remaining players' strategies
- Prediction sometimes called **strategically stable** or **self-enforcing** - no player wants to **unilaterally deviate** from his predicted strategy
- Such a set of strategies is called a **Nash equilibrium**

Nash equilibrium

Definition 2 *In the normal form game G , the strategies s_1^*, \dots, s_N^* are a Nash equilibrium if for each i ,*

$$u_i(s_1^*, \dots, s_i^*, \dots, s_N^*) \geq u_i(s_1^*, \dots, s_i, \dots, s_N^*)$$

for every $s_i \in S_i$; i.e., $s_i^ \in \max_{s_i \in S_i} u_i(s_1^*, \dots, s_i, \dots, s_N^*)$.*

- If this does not hold for for some $i \in \bar{I}$, then a subset of players can profit from deviating
- In the Prisoner's dilemma, both players playing **FINK** is a Nash equilibrium

Battle of the sexes

- Bimatrix game:

		Pat	
		Opera	Baseball
Chris	Opera	(2,1)	(0,0)
	Baseball	(0,0)	(1,2)

- (Opera,Opera) and (Baseball,Baseball) are both Nash equilibria
- Lack of uniqueness

Motivating example: Duopolistic Cournot Model (1838)

- Predates Nash equilibrium ideas
- Firms 1 and 2 produce items q_1 and q_2 ; $Q = q_1 + q_2$
- Let $p(Q)$, the price, be defined as $p(Q) = a - Q$
- Cost of production is $C_i(q_i) = cq_i$ and $c < a$ Normal form:
 - $S_i = [0, \infty)$, $s_i \equiv q_i \geq 0$.
 - $P(Q) = 0, Q > a \implies$ firms will not produce more than a

$$q_1^* = \arg \max_{q_1} q_1(a - (q_1 + q_2^*)) - cq_1$$

$$q_2^* = \arg \max_{q_2} q_2(a - (q_1^* + q_2)) - cq_2$$

- Nash equilibrium in quantities:

$$q_1^* = \frac{1}{2} (a - q_2^* - c)$$

$$q_2^* = \frac{1}{2} (a - q_1^* - c)$$

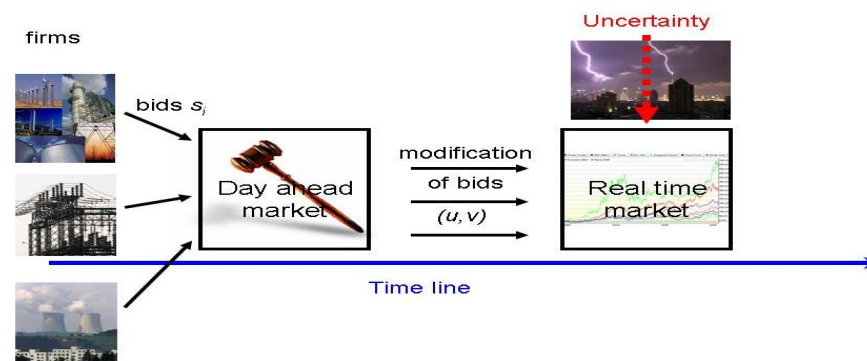
$$q_1^* = q_2^* = \frac{a - c}{3}.$$

- Intuition: If $q_2^* \equiv 0$, $q_1^* = \frac{a-c}{2}$, a monopolist level

Motivating example: Oligopolistic electricity markets

Framework

- Producers have access to a day-ahead market (ignore real-time market)
- Capacity-constrained Cournot bids in day-ahead market
- Variety of price functions for specifying prices



A 2-node Market

- Consider a 2-node electricity market. Suppose node 1 has a demand center while node 2 houses n electricity producers
- Producer i produces q_i MW during a specific hour with capacity C_i
- Producer i 's costs are $c_i q_i + \frac{1}{2} d_i q_i^2$.
- Transmission capacity of the link (1, 2) is infinite
- Price of power at node 1 is given by $p(Q)$, $Q = \sum_{i=1}^N q_i$.

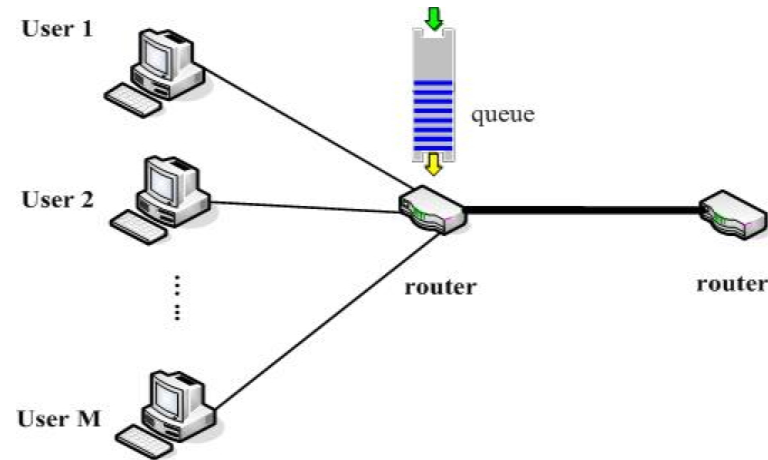
Nash Equilibrium in generation levels

- Producers can affect prices and their payoffs are dependent on the generation levels of the remaining firms
- Specifically, producer i is faced by

$$\begin{array}{ll} G_i(q^{-i}) & \text{maximize}_{q_i} \quad p(Q)q_i - c_iq_i - \frac{1}{2}d_iq_i^2 \\ & \text{subject to} \quad q_i \leq C_i \quad (\lambda_i) \\ & \quad \quad \quad q_i \geq 0, \end{array}$$

where $q^{-i} = (q_j)_{j \neq i}$.

- Nash equilibrium in generation levels: (q_1^*, \dots, q_n^*) where $q_i^* \in \arg \max G_i(q^{*-i})$ for all i .



Flow control games

- Multitude of applications on the internet
- A large number of users - inherently noncooperative in nature
- Demands for bandwidth lead to congestion
- Solution concept - a Nash equilibrium in flow rates, where each agent maximizes its utility less cost of sending flow

- User i has utility given by $U_i(x_i) := \ln(1 + x_i) + d_i$
- Cost of sending flow is $\mathcal{P}(x_i, x^{-i}) := \frac{k_i x_i^2}{N - \sum_{i=1}^m x_i}$.
- Finally, flow decisions are required to stay feasible with respect to the capacity constraint $\sum_{i=1}^N x_i \leq M$.

Nash Equilibrium in flow decisions

- Users maximize the benefit of sending flow by solving

$$\begin{array}{ll} D_i(x^{-i}) & \text{maximize } U_i(x_i) - P(x_i; x^{-i}) \\ & \text{subject to } x_i \geq 0, \end{array}$$

- Nash equilibrium in generation levels: (x_1^*, \dots, x_n^*) where $x_i^* \in \arg \max G_i(x^{-i,*})$, for all i and $\sum_{i=1}^N x_i \leq M$.

Mixed-Strategy Nash Equilibrium

- In a host of practical settings, strategy decisions are continuous variables (as opposed to discrete such as in the bimatrix games).
- Example: Mixed-strategy Nash equilibria
- **Definition 3** *In the normal form game $G = \{S_1, \dots, S_N; u_1, \dots, u_n\}$, let $S_i = \{s_{i1}, \dots, s_{iK}\}$. Then a mixed-strategy for player i is a distribution $\bar{p}_i = (p_{i1}, \dots, p_{iK})$, where $0 \leq p_{ik} \leq 1$ for $k = 1, \dots, K$ and $\sum_{k=1}^K p_{ik} = 1$.*
- Intuitively - a mixed strategy specifies the randomization over the set of pure strategies
- A mixed-strategy Nash equilibrium is given by a set $(\bar{p}_1^*, \dots, \bar{p}_K^*)$.
- Important - mixed-strategies result in continuous decision variables (the probability masses)

Our focus: Continuous Strategy Sets

- This course is devoted to settings where the strategies are continuous variables and payoff functions are twice continuously differentiable.*
- Why? - This allows for casting the resulting Nash equilibrium problem as a related variational inequality/complementarity problem (will be defined shortly).

*We may discuss applications where the differentiability assumption is relaxed.

A Complementarity reformulation

- Consider a Nash equilibrium problem in which player i solves

$$\begin{aligned} (\text{Opt}_i) \quad & \min_{x_i} f_i(x_i; x^{-i}) \\ & x_i \geq 0, \end{aligned}$$

where f_i is player i 's payoff function (convex and twice continuously differentiable).

- The sufficient optimality conditions of each player can be jointly stated as[†]

CP

$$\begin{aligned} 0 \leq x_1 \perp \nabla_{x_1} f_1(x_1; x^{-1}) \geq 0 \\ \vdots \\ 0 \leq x_n \perp \nabla_{x_n} f_n(x_n; x^{-n}) \geq 0, \end{aligned}$$

[†]The symbol \perp implies complementarity; $x \perp y \implies [x]_i [y]_i = 0, \quad \forall i$

or compactly as $0 \leq x \perp F(x) \geq 0$, $F = \begin{pmatrix} \nabla_{x_1} f_1 \\ \vdots \\ \nabla_{x_N} f_N \end{pmatrix}$

- Such a problem is called a **nonlinear complementarity problem** and is defined as

Definition 4 *Given a mapping $F : \mathbb{R}_+^n \rightarrow \mathbb{R}^n$, the $NCP(F)$ is to find a vector $x \in \mathbb{R}_+^n$ satisfying $0 \leq x \perp F(x) \geq 0$.*

A Variational Inequality reformulation

- x^* is a Nash equilibrium if and only if for each player i we have

$$(y_i - x_i)^T \nabla_{x_i} f_i(x_i; x^{-i}) \geq 0, \quad \forall y^i \geq 0.$$

- Concatenating these conditions we have

VI	$(y_1 - x_1)^T \nabla_{x_1} f_1(x_1; x^{-1}) \geq 0,$	$\forall y^1 \geq 0$
	\vdots	
	$(y_N - x_N)^T \nabla_{x_N} f_N(x_N; x^{-N}) \geq 0,$	$\forall y^N \geq 0.$

or equivalently $(y - x)^T F(x) \geq 0, F = \begin{pmatrix} \nabla_{x_1} f_1 \\ \vdots \\ \nabla_{x_N} f_N \end{pmatrix}$

- Such a problem is called a **variational inequality problem** and is defined as follows:

Definition 5 *Given a mapping $F : \mathbb{R}_+^n \rightarrow \mathbb{R}^n$, the $VI(F, \mathbb{R}_+^n)$ is to find a vector $x \in \mathbb{R}_+^n$ satisfying $(y - x)^T F(x) \geq 0$ for all $y \in \mathbb{R}_+^n$.*

A Generalization of Nash

- An important generalization of the Nash solution concept pertains to the strategy sets being related. Specifically, player i solves

$$(GenOpt_i) \quad \min_{x_i \in K(x^{-i})} f_i(x_i; x^{-i})$$

- The resulting variational inequality (called a **quasi variational inequality (QVI)**) is given by

QVI	$(y_1 - x_1)^T \nabla_{x_1} f_1(x_1; x^{-1}) \geq 0,$	$\forall y^1 \in K(x^{-1})$
	\vdots	
	$(y_N - x_N)^T \nabla_{x_N} f_N(x_N; x^{-N}) \geq 0,$	$\forall y^N \in K(x^{-N}).$

Stackelberg Equilibrium Problems

Stackelberg (1934) proposed a dynamic duopolistic model:

- Leader moves first
- Subordinate moves second

Examples:

- Automobile industry: GM is a leader while Ford and Chrysler may be considered to be followers
- Tolling in traffic networks: Network operator sets tolls while users respond and reach an equilibrium subject to tolls

Crucial idea: follower's problem is a **parametrized** VI/CP

Stackelberg-Cournot

Setting:

- Firm 1 chooses quantity $q_1 \geq 0$
- Firm 2 observes q_1 and maximizes $q_2(a - (q_1 + q_2) - c)$.
- Firm 2's best response is $q_2^* = \frac{a - q_1 - c}{2}$
- Firm 1 may then choose q_1 by solving $q_1(a - (q_1 + q_2) - c)$ where q_2 is given by $\frac{a - q_1 - c}{2}$
- This results in $q_1^* = \frac{a - c}{2}$ and $q_2^* = \frac{a - c}{4}$.

More generally

- Suppose firm 1's payoff function is $f(x, y)$
- Firm 2's complementarity problem is $0 \leq y \perp F(y; x) \geq 0$.
- The resulting problem faced by firm 1 is a **mathematical program with complementarity/equilibrium constraints**

MPCC	minimize	$f(x, y)$
	x, y	
	subject to	$0 \leq y \perp F(y; x) \geq 0$.

Why is this difficult

- The constraints are nonsmooth (do not satisfy "regularity" conditions at any feasible point)
- Existing convergence theory for nonlinear programming does not hold

Focus of course

Three broad problem areas:

- Nash equilibrium problems \implies VIs/CPs
- Generalized Nash equilibrium problems \implies QVIs
- Stackelberg equilibrium problems \implies MPECs/MPCCs

Questions in each setting

- Existence/uniqueness of equilibria
- Construction of convergent algorithms for obtaining such equilibria (if they do indeed exist)
 - Global convergence
 - Local convergence

Books

The following books will be adhered to with volume 1 of the first set of references serving as the class text.

- (FP) Facchinei, F. and Pang, J-S, *Finite-dimensional variational inequalities and complementarity problems*, Vols. I and II. Springer Series in Operations Research. Springer-Verlag, New York, 2003
- (CPS) Cottle, R. W. and Pang, J-S and Stone. R. E., *The Linear Complementarity Problem*, Computer Science and Scientific Computing. Academic Press, Inc., Boston, MA, 1992.
- (LPR) Luo, Z-Q, Pang, J-S and Ralph, D., *Mathematical programs with Equilibrium Constraints*, Cambridge University Press, Cambridge, 1996.
- (MWG) Mas-Colell, A., Whinston, M.D. and Green, J.R. *Microeconomic Theory* , Oxford University Press, USA, 1995.
- (B) Bertsekas D. P., *Nonlinear Programming*, Athena Scientific, 1995.
- (BON) Bertsekas Dimitri P., Ozdaglar, A. and Nedic, A., *Convex Optimization and Analysis*, Athena Scientific, 2003.
- (KON) Konnov, I.V. *Equilibrium Models and Variational Inequalities*, Elsevier, 2007.

Lecture Outline - tentative

1. (B,BON) Background: Optimization theory and algorithms (convexity, first and second-order KKT conditions, regularity conditions, duality theory, quick survey of algorithms).
(2)

2. Nash equilibrium models
 - (a) (MWG, KON) Nash equilibrium: Background (1)
 - (b) (FP-I,CPS) Applications and models: Linear complementarity problems (LCPs), nonlinear complementarity problems (NCPs) and variational inequalities (VIs) (2)
 - (c) (FP-I,CPS) Existence and uniqueness of equilibria (3)
 - (d) Algorithms:
 - i. (CPS) Lemke's method (pivot method for LCPs) (1)
 - ii. (FP-II) Interior-point approaches (2)
 - iii. (FP-II) Sequential linearization approaches (1)

3. Generalized Nash equilibrium models

- (a) Applications and models: Quasi-variational inequalities (QVIs) and mixed complementarity problems (mCPs) (2)
- (b) (CPS,FP-I) Existence and uniqueness of equilibria (2)
- (c) Algorithms
 - i. (FP-II) Penalization approaches (2)

4. Stackelberg equilibrium models

- (a) (LPR) Applications and models: Mathematical programs with complementarity constraints (MPCCs) (2)
- (b) (LPR) Strong stationarity and second-order conditions (2)
- (c) Algorithms
 - i. (LPR) Sequential quadratic programming approaches (2)
 - ii. (LPR) Interior-smoothing and Interior-penalty approaches (2)

Grading - really painful

- HW 30%
- Examination 40%
- Project (present some papers) 30%