Cours Bachelier, Paris, France

Game Theory: Fundamentals and Application to Wireless and Electricity Networks (Cours 1 à 4)

Samson Lasaulce L2S (CNRS–CentraleSupelec-Univ. Paris Sud) lasaulce@l2s.centralesupelec.fr

May 18-June 15, 2018

- Start from scratch
- Broad audience
- Overview
- Methodologies
- Application-oriented
- Channels: 1. Slides; 2. Speech; 3. Board

Outline

Class #1:

- 0. Introduction
- 1. Strategic form games

Classes #2 and #3:

2. Dynamic games

- 2.1. Repeated games
- 2.2. Case study 1: Wireless power control
- 2.3. Feasible utility region and Shannon theory
- 2.4. Stochastic games, mean-field games, differential games
- 2.5. Case study 2: Consumption power scheduling

Outline

Class #4:

3. More

- 3.X. Coalition form games
- 3.X. Learning in games
- 3.X. More examples: Game theory and finance?
- 3.X. Case study, viral marketing strategies
- 3.X. Bayesian games, signalling games
- 3.X. Proofs?
- 3.4. Conclusion

Game Theory: Fundamentals and Application to Wireless and Electricity Networks

0. Introduction

Game Theory and Optimization...

Cambridge University, UK, winter 1979



A couple of details about one of the experiments

- 33 ducks.
- Two observers/sites 20 m apart.
- Site 1: 12 items/min.
- Site 2: 24 items/min.

Observations



D. G. Harper, "Competitive foraging in mallards: Ideal free ducks", *Anim. Behav.*, 1982, 30, 575-585.

Ducks become drivers



Stuttgart, Germany, 1969 [Braess 1969]



Input flow = 6			
f(x) = x + 50, g(x) = 10x			
$h(x) = +\infty$	h(x) = x + 10		
83 min	92 min		
$(x_1, x_2) = (3, 3)$	$(x_1', x_2', x_3') = (4, 2, 2)$		

In the real life

- Stuttgart 1969: investments into the road network ⇒ traffic ∑. Section of newly-built road closed ⇒ traffic ↗ [Knödel 1969].
- Seoul 2003: one of the three tunnels shut down to restore a river and a park ⇒ traffic flow improved.

In many other situations: Wireless networks [Cohen and Kelly 1990][Perlaza et al 2009], energy networks [Baillieul et al 2015], ...

Trivial inequality in standard optimization

$$\max_{x \in \mathcal{A}} f(x) \le \max_{x \in \mathcal{B}} f(x)$$

when $\mathcal{A} \subseteq \mathcal{B}$.

This inequality does not hold anymore \leftarrow (partial control + multiple utility functions)

$$x = (x_1, ..., x_K).$$

Why only partial control?

Complexity issues. ① Smart grid example: charging instant selection 48 time-slots and 16 vehicles $48^{16} > 32^{16} > 10^{21}$.



Why only partial control?

Complexity issues. ② Wireless example: channel selection with 16 channels and 16 users $16^{16} = 2^{64} > 10^{18}$.



- Main function decomposition,
- several performance criteria,
- several decision-makers, ...

Distributed optimization (DO): typically about partial control with one DM.

Multi-objective optimization (MOO): typically about one DM with full control + several objectives [Björnson et al 2015].

 "Non-cooperative" game theory (GT): typically about several (virtual/real) DMs with partial control + several objectives.
 What is the meaning of optimality then?

Typical issues in scenarios with partial control and multiple objectives

- Which solution concept to consider as a possible game outcome?
- ► Does it exist for the game of interest? Is it unique?
- ► Is it efficient? How do we measure efficiency? How do we improve it?
- ► NE: What is it? Existence? Uniqueness? Efficiency? Existence of a convergent and implementable algorithm?

Main mathematical representations

- Strategic or normal form games.
- Extensive form games.
- Coalitional form games.

Cheap map of the game theory jungle



[Bacci et al 2016]

Game Theory: Fundamentals and Application to Wireless and Electricity Networks

1. Strategic form games

Game \equiv **triplet**:

$$\mathcal{G} = (\mathcal{K}, \{\mathcal{S}_i\}_{i \in \mathcal{K}}, \{u_i\}_{i \in \mathcal{K}}).$$

▶ $\mathcal{K} = \{1, ..., K\}$ is the set of players.

- ▶ S_i is the set of strategies for player *i*.
- ► Player *i*'s payoff/utility function:

$$u_{i}: \mathcal{S}_{1} \times ... \times \mathcal{S}_{i} \times ... \times \mathcal{S}_{K} \rightarrow \mathbb{R}$$

$$\underbrace{(s_{1}, ..., s_{i}, ..., s_{K})}_{s: \text{ strategy profile}} \mapsto u_{i}(s_{i}, s_{-i}) .$$

Remark on the strategic form

 \exists a more general form:

$$\mathcal{G} = (\mathcal{K}, \{\mathcal{S}_i\}_{i \in \mathcal{K}}, \succeq_i)$$

Proposition (Debreu 1954): there exist no utility functions for lexicographic ordering on \mathbb{R}^2 .

Proposition: there exists a utility function for every transitive and complete ordering on any countable set:

- completeness: $x \succeq y$ or $y \succeq x$ or both;
- transitivity: " $x \succeq y$ and $y \succeq z$ " $\Rightarrow x \succeq z$.

Proposition (Debreu 1954): there exists a utility function for every transitive, complete, and continuous ordering on a continuous set $\mathcal{X} \subset \mathbb{R}^N$ provided \mathcal{X} is non-empty, closed, and connected:

• continuity: $\mathcal{B}(x) = \{y \in \mathcal{X} : x \succeq y\}$ and $\mathcal{W}(x) = \{y \in \mathcal{X} : y \succeq x\}$ are closed.

Remark (connectedness): \mathcal{X} is said to be disconnected if it is the union of two disjoint nonempty open sets. Otherwise, \mathcal{X} is said to be connected.

Theorem (preferences over lotteries): the complete and transitive preference ordering \succeq over $\Delta(S)$ admits a utility function (expected utility) if and only if \succeq meets the VNM axioms of independence and continuity:

- VNM independence axiom: $x \succ y \Rightarrow (1 \mu)x + \mu z \succ (1 \mu)y + \mu z, \ \mu \in]0,1[;$
- VNM continuity axiom: $x \succ y \succ z \Rightarrow \exists \mu \in]0, 1[, (1-\mu)x + \mu z \succ y \succ (1-\mu)z + \mu x.$

Remarks: the Allais paradox (1953), voting procedures.

Market power terminology [Singh 2009]:

▶ Players
$$\mathcal{K} = \{G1, G2\}.$$

- ► Strategies are merely actions $S_1 = S_2 = {$ low, high $}.$
- ► Utility function for Player 1:

$$u_{1}(s_{1}, s_{2}) = \begin{vmatrix} 0 & \text{if} & (s_{1}, s_{2}) &= & (\text{low}, \text{high}) \\ 1 & \text{if} & (s_{1}, s_{2}) &= & (\text{high}, \text{high}) \\ 3 & \text{if} & (s_{1}, s_{2}) &= & (\text{low}, \text{low}) \\ 4 & \text{if} & (s_{1}, s_{2}) &= & (\text{high}, \text{low}) \end{vmatrix}$$

G1, G2	high price	low price
high price	(3, 3)	(0, 4)
low price	(4, 0)	(1, 1)

A fundamental solution concept: The Nash equilibrium (NE)

Pure Nash equilibrium. Strategy vector/profile such that

$$\forall i \in \mathcal{K}, \forall s_i \in \mathcal{S}_i, \ u_i(s_i^*, s_{-i}^*) \ge u_i(s_i, s_{-i}^*).$$

Mixed Nash equilibrium ...

Mixed strategies



Mixed strategies



• Mixed strategies
$$\pi_i \in \Delta(\mathcal{S}_i)$$
 with
 $\Delta(\mathcal{S}_i) = \left\{ x \in \mathbb{R}^{|\mathcal{S}_i|} : x_j \ge 0, \sum_j x_j = 1 \right\}$

Expected utility

$$\widetilde{u}_i(\pi_1, ..., \pi_K) = \mathbb{E}_{\pi_1 \otimes ... \otimes \pi_K} \left[u_i(s_1, ..., s_K) \right].$$

Mixed Nash equilibrium

$$\forall i \in \mathcal{K}, \forall \pi_i \in \Delta(\mathcal{S}_i), \ \widetilde{u}_i(\pi_i^*, \pi_{-i}^*) \ge \widetilde{u}_i(\pi_i, \pi_{-i}^*).$$

► Stability property (once you are there).

Dynamical property (to get there).

► It "always" exists.

Best-response

$$BR_i(s_{-i}) = \arg\max_{s_i \in S_i} u_i(s_i, s_{-i}).$$

Illustration for continuous sets [Cournot 1838]



The sequential best-response dynamics (1/3)

G1, G2	high	low
high	(3, 3)	(0, 4)
low	(4, 0)	(1, 1)

The sequential best-response dynamics (2/3)

G1, G2	high	low
high	(3, 3)	(0, 4)
low	(4, 0)	(1, 1)
The sequential best-response dynamics (3/3)

G1, G2	high	low
high	(3, 3)	(0, 4)
low	(4, 0)	(1, 1)

A strategy profile s^* is an NE of G iff:

$$s_i^* \in \mathrm{BR}_i(s_{-i}^*) \iff s^* \in \mathrm{BR}(s^*)$$

where

BR: $\mathcal{S} \rightarrow \mathcal{S}$ $s \mapsto BR_1(s_{-1}) \times BR_2(s_{-2}) \times ... \times BR_K(s_{-K})$

Nash existence theorem [Nash 1950].

 $S = S_1 \times ... \times S_K$ is finite. Then, there is a <u>mixed</u> NE.

Kuhn existence theorem [Kuhn 1953]. Every

finite game of perfect information has at least one pure NE.

Glicksberg theorem [Glicksberg 1952]. S_i compact, u_i continuous in s. Then, there is a <u>mixed</u> NE.

Debreu-Fan-Glicksberg theorem [Debreu, Fan, Glicksberg 1952]. Above assumptions & u_i quasiconcave in s_i . Then, there exists a pure NE.

More about the existence of NE



[Lasaulce & Tembine 2011]

Static games	Dynamic games
Existence	Existence
Uniqueness	Utility region characterization ; uniqueness
Efficiency	Design of strategies

Rosen theorem [Rosen 1965]

- ► S_i compact convex.
- ► u_i continuous in s.
- ▶ u_i concave in s_i .
- ► Diagonally strict concavity:

$$\exists r > 0, \ \forall s \neq s', \ [s'-s]^T [\gamma_r(s) - \gamma_r(s')] > 0$$

where

$$\gamma_r(s) = \left(r_1 \frac{\partial u_1}{\partial s_1}(s), ..., r_K \frac{\partial u_K}{\partial s_K}(s) \right).$$

Then, there is a unique pure NE.

Concave game example

Utility:

$$u_1(\boldsymbol{A}_1, \boldsymbol{A}_2) = \mathbb{E} \log \left| \boldsymbol{I} + \boldsymbol{X}_1 \boldsymbol{A}_1 \boldsymbol{X}_1^H + \boldsymbol{X}_2 \boldsymbol{A}_2 \boldsymbol{X}_2^H \right| - \mathbb{E} \log \left| \boldsymbol{I} + \boldsymbol{X}_2 \boldsymbol{A}_2 \boldsymbol{X}_2^H \right|$$

Action space:

$$\mathcal{A}_1 = \left\{ \boldsymbol{A}_1 \ge 0, \boldsymbol{A}_1^H = \boldsymbol{A}_1, \mathrm{Tr} \boldsymbol{A}_1 \le a \right\}$$

1. DSC is met (trace inequality); 2. NE determination (random matrix theory) [Belmega et al 2011]

Definition (standard functions) A vector function $g : \mathbb{R}^K_+ \to \mathbb{R}^K_+$ is standard if we have:

▶ Monotonicity: $\forall (x, x') \in \mathbb{R}^{2K}_+$, $x \leq x' \Rightarrow g(x) \leq g(x')$.

► Scalability: $\forall \alpha > 1, \forall x \in \mathbb{R}_+^K, g(\alpha x) < \alpha g(x).$

Theorem [Yates 1995] If $BR = (BR_1, ..., BR_K)$ is standard, then there is a unique pure NE.

Remark: BR intersection.

More about the uniqueness of NE



[Lasaulce & Tembine 2011]

- Efficiency: typical consequence of partial control
- Correlation: mixed NE assume independent lotteries
- Strategic stability: only stable to single deviations
- Not fully adapted to QoS constraints

For more drawbacks see [Perlaza & Lasaulce 2014]

Solution concepts for strategic/extensive form games

- Pure/mixed Nash equilibrium, Wardrop equilibrium,
- correlated equilibrium, coarse correlated equilibrium,
- N- strong equilibrium,
- Nash equilibrium refinements : trembling hand perfect equilibrium, proper equilibrium,
- ϵ -Nash equilibrium,
- logit equilibrium,

Solution concepts for strategic/extensive form games. Continued

- maxmin strategy profiles,
- Bayesian equilibrium,
- evolutionary stable solution,
- satisfaction equilibrium, generalized Nash equilibrium,
- Stackelberg equilibrium,
- Pareto optimum, social optimum,
- bargaining solutions (Nash, egalitarian, Kalai-Smorodinsky, etc.),...

Definition (Pareto-dominance): s Pareto-dominates s' if:

$$\forall i \in \mathcal{K}, \ u_i(s) \ge u_i(s'),$$

with strict inequality for at least one player.

Definition (Pareto-optimum): s^* is Pareto-optimal (-efficient) if it is dominated by no other profile.

Illustration of Pareto optimality



Definition (social welfare): the social welfare of a game is defined as:

$$w = \sum_{i=1}^{K} u_i.$$

Definition (social optimum): an SO is a strategy profile which maximizes w.

Remark: An SO is a PO.

Definition (price of anarchy):

$$\mathrm{PoA} = \frac{\max_{s \in \mathcal{S}} w(s)}{\min_{s^* \in \mathcal{S}^{\mathrm{NE}}} w(s^*)}$$

where $\mathcal{S}^{\rm NE}$ is the set of NE of the game.

Definition (price of stability):

$$\operatorname{PoS} = \frac{\max_{s \in \mathcal{S}} w(s)}{\max_{s^* \in \mathcal{S}^{\operatorname{NE}}} w(s^*)}.$$

[Papadimitriou 2001] [Anshelevich et al 2004].

Example: PoA in non-atomic routing games

The network cost is defined by:



Theorem. For polynomials costs of maximum degree d, the PoA is bounded as:

degree	1	2	3	4		d
PoA	$\frac{4}{3}$	1.626	1.896	2.151	•••	$\Omega(\frac{d}{\ln(d)})$

[Correa et al 2005].

Possible approaches (non-exhaustive list)

- ► Introduce pricing.
- ► Introduce hierarchy.
- ► Introduce coordination (e.g., correlated equilibrium).
- ► Introduce cooperation (bargaining, cooperation plan in dynamic games, agreement/contract in coalitional games, ...).

How to improve efficiency (pricing). Example



Utility of player k when connecting to base station s

$$\upsilon_{k,s}(x) = \log\left[1 + \frac{1}{a_s + bx_s}\right],$$

 $a_s > 0, b > 0.$

Example (illustration)

Social welfare for S = 2



Example (pricing and modified game)

Let n_k be the data volume to be transferred:

$$\tau_{k,s}(x) = \frac{n_k}{v_{k,s}(x)}.$$

Cost function of the new game:

$$c_{k,s}(x) = p(\tau_{k,s}(x)) + \beta_s.$$

Parameter adjustment \rightarrow desired solution.

How to improve efficiency: introduce coordination through correlated equilibria

Definition (correlated equilibrium) Let $\sigma_k : \mathcal{A}_k \to \mathcal{A}_k$ be a mapping. Then q^{CE} is a CE if

$$\sum_{a \in \mathcal{A}} q^{\mathrm{CE}}(a_k, a_{-k}) u_k(a_k, a_{-k}) \ge \sum_{a \in \mathcal{A}} q^{\mathrm{CE}}(a_k, a_{-k}) u_k(\sigma_k(a_k), a_{-k}),$$

Example (CR coordination game)

	Low	High
High	(5,1)	(0, 0)
Low	(4, 4)	(1,5)

Set of correlated equilibria



Definition The NBS is the unique solution of

$$\max_{\substack{(u_1,u_2) \in \mathcal{U} \\ \text{subject to}}} (u_1 - \lambda_1)(u_2 - \lambda_2)$$
$$u_1 \ge \lambda_1, u_2 \ge \lambda_2$$

where \mathcal{U} is the game feasible utility set.

Illustration of the NBS



[FIG7] The graphical interpretation of the NBS point (red circle) as the intersection between the Pareto boundary of \mathcal{U} and the hyperbola $(u_1 - \lambda_1)(u_2 - \lambda_2) = \kappa$, where the status quo $\lambda = (\lambda_1, \lambda_2)$ is represented by the blue diamond.

Game Theory: Fundamentals and Application to Wireless and Electricity Networks

2. Dynamic games

Typical ingredients

- ► Several stages.
- ► Notions of game history, action plans.
- ► Average/long-term utility.
- ▶ The stage utility is state-dependent $(u_i(a, x))$.

Informal definition. A game in which at least one player can use a strategy depending on previously played actions. No universal definition, only special classes.

- ► Repeated games (∅).
- ► Stochastic games (MDP).
- ► Differential/difference games (OC).
- ► Mean-field games.

► Evolutionary games.

Repeated games with perfect monitoring

Definition (game history): $\forall t \geq 1$, $h_t = (a(1), ..., a(t-1)) \in \mathcal{H}_t$ where $\mathcal{H}_t = \mathcal{A}^{t-1}$.

Definition (pure strategy): A pure strategy for player $i \in \mathcal{K}$ is a sequence $(\tau_{i,t})_{t\geq 1}$ with

$$\begin{aligned}
\tau_{i,t} : & \mathcal{H}_t \to \mathcal{A}_i \\
& h_t \mapsto a_i(t)
\end{aligned}$$

Definition (behavior strategy): A behavior strategy for player $i \in \mathcal{K}$ is a sequence $(\tilde{\tau}_{i,t})_{t>1}$ with

$$\widetilde{\tau}_{i,t} : \begin{array}{ccc} \mathcal{H}_t & \to & \Delta(\mathcal{A}_i) \\ & h_t & \mapsto & \pi_i(t). \end{array}$$

Repeated games utilities

Finitely repeated games. Let $\tau = (\tau_1, ..., \tau_K)$ and $T \ge 1$:

$$v_i^T(\tau) = \frac{1}{T} \sum_{t=1}^T u_i(a(t)).$$

Infinitely repeated games:

$$v_i^{\infty}(\tau) = \lim_{T \to +\infty} \frac{1}{T} \sum_{t=1}^T u_i(a(t)).$$

Discounted repeated games. Let $0 < \lambda < 1$ be the discount factor:

$$v_i^{\lambda}(\tau) = \sum_{t=1}^{+\infty} \lambda (1-\lambda)^{t-1} u_i(a(t)).$$

Definition (equilibrium strategies). A joint strategy τ^* supports an equilibrium of the repeated game $(\mathcal{K}, \{\mathcal{T}_i\}_{i \in \mathcal{K}}, \{v_i^y\}_{i \in \mathcal{K}}), y \in \{T, \infty, \lambda\}, \text{ if:}$

 $\forall i \in \mathcal{K}, \forall \tau'_i, \ v^y_i(\tau^*) \ge v^y_i(\tau'_i, \tau^*_{-i}).$

Remark (equilibrium analysis): Existence for finite games, compact games, static games with a Nash equilibrium. In contrast with static games, there can be many equilibria.

Equilibrium characterization for discounted repeated games with perfect monitoring

Folk theorem. The set of equilibrium <u>utilities</u> when $\lambda \to 0$ is given by

$$E_0 = \operatorname{IR}(\mathcal{G}) \cap \operatorname{co}\left(\mathcal{U}(\mathcal{G})\right)$$

where:

• IR(
$$\mathcal{G}$$
) = { $u \in \mathbb{R}^K, \forall i \in \mathcal{K}, u_i \ge \min_{\pi_{-i}} \max_{\pi_i} \widetilde{u}_i(\pi)$ };

• $\mathcal{U}(\mathcal{G}) = \{ u' \in \mathbb{R}^K : \exists a, u(a) = u' \}.$

Repeated prisoner's dilemma



Relaxing the perfect monitoring assumption: 2-connected graphs

Definition (strongly connected graph) A graph Γ is said to be strongly connected if for each pair of vertices (i, j), there is a directed path from *i* to *j*.

Definition (2-connected graph) The graph Γ is 2connected if, for any vertex *i*, $\Gamma \setminus \{i\}$ is strongly connected.

Theorem The following two assertions are equivalent:

(i) the observation graph of the infinitely repeated games is 2-connected;

(ii) $E_{\infty} = \operatorname{IR}(\mathcal{G}) \cap \operatorname{co}(\mathcal{U}(\mathcal{G})).$

[Renault and Tomala 1998]
- ► Static game formulation.
- ► A repeated game formulation.

Near far effect



A wireless power control scenario



Modeling the problem as a static game [Goodman & Mandayam 2000]

Time slots



Modeling the problem as a static game. Continued

- ► Set of players : $\mathcal{K} = \{1, ..., K\}$.
- ► Set of actions : $\mathcal{A}_i = [0, A^{\max}].$
- Utilities : energy-efficiency;

$$u_i(a_i, a_{-i}) = \frac{\text{benefit}}{\text{cost}} = \frac{f(\beta_i)}{a_i}$$

where

$$\beta_i = \frac{g_i a_i}{1 + \sum_{j \neq i} g_j a_j}.$$

Properties assumed for f



- f non-negative, continuous, and non-decreasing.
- f sigmoidal.

•
$$\lim_{x \to 0} \frac{f(x)}{x} = 0$$
, $\lim_{x \to +\infty} f(x) = const \le 1, 0 \le f(x) \le 1$.

Nash equilibrium analysis (1/3)

Existence

Existence

- $\mathcal{A}_i = [0, A_i^{\max}]$: compact, convex.
- u_i is continuous w.r.t. $a = (a_1, ..., a_K)$.
- u_i is quasi-concave w.r.t. a_i (f(x) sigmoidal $\Rightarrow \frac{f(x)}{x}$ is quasi-concave).

Nash equilibrium analysis (2/3)

Uniqueness

Uniqueness

The best response is a function and

$$\forall i \in \mathcal{K}, \ BR_i(a_{-i}) = \frac{\beta}{g_i} \left(1 + \sum_{j \neq i} g_j a_j \right)$$

with
$$\beta^* f'(\beta) = f(\beta)$$
.

The game is standard:

- Monotonicity: $a' \leq a \Rightarrow BR(a') \leq BR(a)$.
- Scalability: $\forall \alpha > 1$, $BR(\alpha a) < \alpha BR(a)$.

Determination (interior point)

Solve the system of equations $\frac{\partial u_i}{\partial a_i}(a) = 0$, which leads to:

$$\forall i \in \{1, ..., K\}, \ a_i^{\star} = \frac{1}{g_i} \frac{\beta}{1 - (K - 1)\beta}.$$

Problem Generally inefficient solution. How to improve

efficiency?

Main points

► New utility:

$$\widetilde{u}_i(a) = u_i(a) - \alpha a_i, \ \alpha \ge 0.$$

▶ Good news. The new NE profile Pareto-dominates a^* .

Bad news. Uniqueness not guaranteed, convergence under some specific assumption. Global state information is required.

[Saraydar et al 2002].

Strategic form

$$\mathcal{G}^m = (\mathcal{K}, \{\mathcal{T}_i\}_i, \{v_i^m\}_i) \text{ with } m \in \{T, \lambda\}.$$

If m = T:

$$v_i^T = \frac{1}{T} \sum_{t=1}^T u_i(\underline{a}(t)).$$

If $m = \lambda \in (0, 1]$:

$$v_i^{\lambda} = \sum_{t=1}^{+\infty} \lambda (1-\lambda)^{t-1} u_i(\underline{a}(t)).$$

[Le Treust and Lasaulce 2010]

Public signal choice

$$\omega(t) \triangleq 1 + \sum_{i=1}^{K} g_i a_i(t) = a_i(t)g_i \times \frac{\beta_i(t) + 1}{\beta_i(t)}$$

Pure strategies

$$\tau_{i,t}: \begin{array}{ccc} \left(\mathcal{A}_i \times \Omega\right)^{t-1} & \to & \left[0, A_i^{\max}\right] \\ \left(a_i^{t-1}, \omega^{t-1}\right) & \mapsto & a_i(t) \end{array}$$

.

where

•
$$a_i^{t-1} = (a_i(1), a_i(2), ..., a_i(t-1));$$

• $\omega^{t-1} = (\omega(1), \omega(2), ..., \omega(t-1));$
• $\Omega = \left[1, 1 + \sum_{i=1}^{K} g_i^{\max} A_i^{\max}\right].$

An interesting Nash equilibrium of \mathcal{G}^m , m = T

Proposed equilibrium point

$$\tau_{i,t}^* = \begin{vmatrix} a_i^{\text{OP}} & \text{if } t \in \{1, 2, \dots, T - t_0\} \\ a_i^* & \text{if } t \in \{T - t_0 + 1, \dots, T\} \\ A_i^{\text{max}} & \text{if } \end{vmatrix} \text{ and } \omega(t) = \frac{1 - \gamma}{1 - (K - 1)\gamma} \\ \omega(t) \neq \frac{1 - \gamma}{1 - (K - 1)\gamma} \\ \omega(t) \neq \frac{1 - \gamma}{1 - (K - 1)\gamma} \\ \omega(t) \neq \frac{1 - \gamma}{1 - (K - 1)\gamma} \end{vmatrix}$$

where $\gamma [1 - (K - 1)\gamma] f'(\gamma) - f(\gamma) = 0$ and

$$\forall i \in \mathcal{K}, \ p_i^{\text{OP}} = \frac{1}{g_i} \frac{\gamma}{1 - (K - 1)\gamma}.$$

Comments

- ► To obtain OP, impose $g_j a_j = \text{const.}$
- \blacktriangleright *t*⁰ comes from the equilibrium condition:

Let
$$t_0 = \begin{bmatrix} \frac{f(\alpha)}{\alpha} - \frac{f(\beta)[1 - (K-1)\beta]}{\beta} \\ \frac{f(\alpha)[1 - (K-1)\alpha]}{\alpha} - \frac{f(\alpha)}{\alpha(1 + \sum_{j \neq i} g_j P_j^{\max})} \end{bmatrix}$$

► Local knowledge. Pareto domination of the NE of *G*. Good in terms of social welfare.

► Repeated game methodology holds (worst-case scenario). For instance, t_0 becomes:



▶ Stochastic game formulation: i.i.d. state, $\overline{v}_i^T = \mathbb{E}_g \left[v_i^T(.) \right]$

- good: better performance;
- bad: more information is needed (parameter distribution).

Illustration (fixed parameter)



Illustration (time-varying parameter)



[Mériaux et al 2011]

- Stochastic game case = most general case + most efficient policies.
- ► Importance of characterizing equilibrium points.

Reminders

Feasible set characterization for stochastic games with i.i.d. common state

- ► Stage utilities: $u_i(a_0, a_1, ..., a_K)$; $a_i \in A_i$, $|A_i| < \infty$
- ► Observation/signal structure: $\exists (s_i | a_0), \Gamma(y_i | a_0, a_1, ..., a_K);$ $|S_i| < \infty, |Y_i| < \infty$
- ► Long-term utilities:

$$v_i^{\infty}(\tau_1, ..., \tau_K) = \lim_{T \to +\infty} \frac{1}{T} \sum_{t=1}^T \mathbb{E} \left[u_i(A_0(t), A_1(t), ..., A_K(t)) \right]$$

References: [Larrousse and Lasaulce 2013][Larrousse et al 2015][Larrousse et al 2018]

Stage game description (example)

▶ Decision-makers: $\{1, 2\}$; $0 \equiv$ nature.

► Action sets:
$$A_0 = A_1 = A_2 = \{0, 1\}.$$

► Stage utility function:

$$u(a_0, a_1, a_2) = \begin{vmatrix} 1 & \text{if } a_0 = a_1 = a_2 \\ 0 & \text{otherwise} \end{vmatrix}$$

٠

Observation structure

▶ Stages: $t \in \{1, 2, ..., T\}$, $T \ge 2$.

▶ DM 1 knows $a_0^T = (a_0(1), a_0(2), ..., a_0(T))$ and has perfect recall.

► $\forall t \ge 2$, DM 2 knows $a_0^t = (a_0(1), ..., a_0(t-1))$, perfectly monitors DM 1's actions $a_1^t = (a_1(1), ..., a_1(t-1))$, and has perfect recall.

Question: To what extent can they coordinate?

$$\mathbb{E}\left[\frac{1}{T}\sum_{t=1}^{T}u(A_0(t),A_1(t),A_2(t))\right] = \mathbb{E}\left[\frac{1}{T}\sum_{t=1}^{T}u_t\right]$$

Trivial upper bound

Centralized case

$$\mathbb{E}\left[\frac{1}{T}\sum_{t=1}^{T}u_{t}\right] \leq \mathbb{E}\left[\frac{1}{T}\sum_{t=1}^{T}\max_{(a_{1},a_{2})}u(a_{0}(t),a_{1},a_{2})\right] = 1.$$

Average utility

► Scheme 1:



► Average utility

$$\mathbb{E}\left[\frac{1}{T}\sum_{t=1}^{T}u_t\right] \to \frac{1}{2} = 0.5. \quad \text{ for } A_0 \sim \mathcal{B}\left(\frac{1}{2}\right)$$

Average utility



► Average utility:

$$\mathbb{E}\left[\frac{1}{T}\sum_{t=1}^{T}u_t\right] \to \frac{5}{8} = 0.625.$$

Maximal average utility

$$\mathbb{E}\left[\frac{1}{T}\sum_{t=1}^{T}u_t\right] \to \gamma^* \simeq 0.81$$

where

$$\gamma^{\star}$$
 is the solution of $\frac{h(x) - 1}{x - 1} = \log_2 3$

and $h(x) = -x \log_2 x - (1-x) \log_2 (1-x)$.

Strategies

► Causal case:

$$\tau_{i,t}: \qquad \mathcal{S}_i^t \times \mathcal{Y}_i^{t-1} \qquad \to \quad \mathcal{A}_i \\ (s_i(1), ..., s_i(t), y_i(1), ..., y_i(t-1)) \qquad \mapsto \quad a_i(t)$$

► Noncausal case:

$$\tau_{i,t}: \qquad \mathcal{S}_i^T \times \mathcal{Y}_i^{t-1} \qquad \to \quad \mathcal{A}_i \\ (s_i(1), ..., s_i(\mathbf{T}), y_i(1), ..., y_i(t-1)) \qquad \mapsto \quad a_i(t)$$

Important observation

Important observation

$$\begin{aligned} v_i^{\infty}(\tau_1, ..., \tau_K) \\ &= \lim_{T \to +\infty} \frac{1}{T} \sum_{t=1}^T \mathbb{E} \left[u_i(A_0(t), A_1(t), ..., A_K(t)) \right] \\ &= \lim_{T \to +\infty} \frac{1}{T} \sum_{t=1}^T \sum_{a_0, ..., a_K} P_{A_0(t), ..., A_K(t)}(a_0, ..., a_K) u_i(a_0, ..., a_K) \\ &= \sum_{a_0, ..., a_K} u_i(a_0, ..., a_K) \lim_{T \to +\infty} \frac{1}{T} \sum_{t=1}^T P_{A_0(t), ..., A_K(t)}(a_0, ..., a_K) \end{aligned}$$

Definition $Q(a_0, a_1, ..., a_K)$ is implementable if $\exists (\tau_1, ..., \tau_K)$ s.t.

$$\frac{1}{T} \sum_{t=1}^{T} P_{A_0(t),\dots,A_K(t)}(a_0,\dots,a_K) \to Q(a_0,a_1,\dots,a_K)$$

Characterization of implementable distributions (noncausal case)

Theorem 1

- $(A_0(t))_{t \ge 1} \text{ i.i.d; } A_0 \sim \rho_0$ K = 2
- $-A_1(t) = \tau_{1,t}(A_0(1), ..., A_0(T))$ $-A_2(t) = \tau_{2,t}(A_1(1), ..., A_1(t-1))$
- Then $Q(a_0, a_1, a_2)$ is implementable iff its marginal w.r.t (a_1, a_2) is ρ_0 and

 $H_Q(A_0, A_1, A_2) \ge H_Q(A_0) + H_Q(A_2).$

$$\begin{array}{lll} \text{minimize} & -\sum\limits_{a_0,a_1,a_2} Q(a_0,a_1,a_2) w(a_0,a_1,a_2) \\ \text{subject to} & H_Q(A_0) + H_Q(A_2) - H_Q(A_0,A_1,A_2) &\leq 0 \\ & -Q(a_0,a_1,a_2) &\leq 0 \\ & -1 + \sum\limits_{a_0,a_1,a_2} Q(a_0,a_1,a_2) &= 0 \\ & -\rho_0(a_0) + \sum\limits_{a_1,a_2} Q(a_0,a_1,a_2) &= 0 \end{array}$$

► General case [Larrousse et al ITW 2015]

$$I_Q(S_1; A_2) \le I_Q(V; Y_2|A_2) + I_Q(V; S_1|A_2)$$

where auxiliary variables are used.
Characterization of implementable distributions (causal case)

Theorem 2

- $-(A_0(t))_{t\geq 1}$ i.i.d. + memoryless O.S. - $K\geq 2$
- $-A_i(t) = \tau_{1,t}(S_i(1), \dots, S_i(t), Y_i(1), \dots, Y_i(t-1))$
- ► Then $Q(a_0, ..., a_K)$ is implementable iff it factorizes as

$$Q(a_0, ..., a_K) = \sum_{z, s_1, ..., s_k} \rho_0(a_0) \exists (s_1, ..., s_k | a_0) P_Z(z) \prod_{k=1}^K P_{A_k | S_k, Z}(a_k | s_k, z)$$

[Larrousse et al 2015][Gossner et al 2006]

- Security aspect
- Continuous case
- Controlled state

Definition (Stochastic games with individual states): a stochastic game with individual states is a 6-uplet $\mathcal{G} = \left(\mathcal{K}, \{\mathcal{A}_i\}_i, \{\mathcal{X}_i\}_i, \{\mathcal{A}_i\}_i, \{\mathcal{A}_i\}_i, \{\alpha_i\}_i, q, \{u_i\}_i\right)$ where

- Ω_i is the set of individual states of player *i*;
- $\widetilde{\mathcal{A}}_i(x_i)$ is the set of feasible actions for the state $x_i \in \mathcal{X}_i$;
- $\alpha_i : \mathcal{X}_i \to 2^{\mathcal{A}_i}$ is the correspondence determining the feasible actions at a given state of the game;
- under the Markov game assumption, the transition probability of the states is given by:

$$q: \left| \begin{array}{ccc} \mathcal{X} \times \bigotimes_{i=1}^{K} 2^{\mathcal{A}_{i}} & \to & \Delta(\mathcal{X}) \\ (\underline{x}, \underline{a}) & \mapsto & q(\underline{x}' | \underline{x}, \underline{a}) \end{array} \right|$$

with $\mathcal{X} = \mathcal{X}_1 \times \ldots \times \mathcal{X}_K$.

Equilibrium analysis for stochastic games

- Common state with perfect monitoring and recall + finite action and state spaces: there exists an equilibrium in the finitely/discounted repeated games (see Shapley 1953 for 2-player zero-sum games and Takahashi 1962 Fink 1964 for non-zero-sum games).
- Individual states with perfect monitoring and recall: there exists an equilibrium in the finitely/discounted repeated games (see Vrieze 2007).
- Common state + perfect monitoring + irreducible stochastic games: there is a Folk theorem for infinitely repeated games (Dutta 1991).
- Common state + public signal + irreducible stochastic games: there is a Folk theorem for infinitely repeated games (Hörner etal 2009, Fudenberg and Yamamoto 2009).

• ...

Differential games (linear-quadratic + common state + finite horizon)

- Control functions: $u_i : t \mapsto u_i(t)$, $i \in \{1, ..., K\}$
- State law:

$$\frac{\mathrm{d}x}{\mathrm{d}t}(t) = \mathbf{A}(t)x(t) + \sum_{i=1}^{K} \mathbf{B}_{i}(t)u_{i}(t)$$

• Cumulative utility:

$$J_i(u_1, ..., u_K) = \int_{t \in [0,T]} x^{\mathrm{T}}(t) \mathbf{Q}_i x(t) \mathrm{d}t + \sum j = 1^K \int_{t \in [0,T]} u_j^{\mathrm{T}}(t) \mathbf{R}_{ij} u_j(t) \mathrm{d}t + q_i(x_T) \mathrm{d}t$$

More general differential games

• More general control law:

 $u_i(t, y_i(t))$

• More general state law:

$$\frac{\mathrm{d}x}{\mathrm{d}t}(t) = f(t, x(t), u_1(t, y_1(t)), \dots, u_K(t, y_K(t)))$$

- More general observation structures. Closed-loop perfect state example: $y_i(t) = \{x(t') : 0 \le t' \le t\}$. Memoryless perfect state example: $y_i(t) = \{x(0), x(t)\}$.
- Remark (stochastic differential game):

 $dx(t) = f(t, x(t), u_1(t, y_1(t)), \dots, u_K(t, y_K(t)))dt + dw(t)$

 \rightarrow One path to mean field games.

Equilibrium analysis for differential/difference games

Basar, T., Olsder, G.J., 1982. Dynamic noncooperative game theory. In: Classics in Applied Mathematics, first ed. SIAM, Philadelphia.

T. Eisele, "Nonexistence and Nonuniqueness of Open-Loop Equilibria in Linear-Quadratic Differential Games", Journal of optimization theory and applications, Vol. 37, No. 4, Aug. 1982.

G. Jank, "Introduction to Non-cooperative Dynamical Game Theory", Coimbra, March 2001.

G. Jank and H. Abou-Kandil, "Existence and Uniqueness of Open-Loop Nash Equilibria in Linear-Quadratic Discrete Time Games", IEEE Trans. on Automatic Control, Vol. 48, No 2, Feb. 2003.

Engwerda, J.C., 2005. LQ Dynamic Optimization and Differential Games. Wiley.

P. Cardaliaguet, "Introduction to differential games", Lecture Notes, 2010.

M. Quincampoix, "Differential games", Computational complexity. Vols. 16, 854861, Springer, New York, 2012.

Game Theory: Fundamentals and Application to Wireless and Electricity Networks

3. Learning algorithms and strategic-form games

Updating rule (asynchronous BRD) K = 2. Action sequence: $a_1(0)$, $a_2(1) \in BR_2[a_1(0)]$, $a_1(2) \in BR_1[a_2(1)]$, etc. More generally:

$$a_i(t+1) \in BR_i [a_1(t+1), ..., a_{i-1}(t+1), a_{i+1}(t), ..., a_K(t)].$$

Updating rule (synchronous BRD):

$$a_i(t+1) \in BR_i[a_{-i}(t)].$$

[Cournot 1838]

Main features

- ► Fast convergence.
- ► Steady state: NE.

► Required knowledge: Action profile and individual utility function (in general).

The iterative water-filling algorithm [Yu et al 2002]

► Actions:
$$a_i = p_i = (p_{i,1}, ..., p_{i,S})$$
 with $\sum_s p_{i,s} \le P^{\max}$ and $p_{i,s} \ge 0$

► BRD:

$$p_i(t+1) \in \operatorname*{argmax}_{p_i} \sum_{s=1}^{S} \log \left(1 + \frac{g_{ii,s} p_{i,s}}{\sigma^2 + \sum_{j \neq i} g_{ji,s} p_{j,s}(t)} \right)$$

► The water-filling solution writes as

$$p_{i,s}(t+1) = \left[\frac{1}{\lambda_i} - \frac{p_i(t)}{\text{SINR}_i(t)}\right]^+$$

Algorithm 2: Fictitious play (FP)

Updating rule (synchronous FP):

$$a_i(t+1) \in \arg \max_{a_i \in \mathcal{A}_i} \sum_{a_{-i}} f_{-i,t}(a_{-i}) u_i(a_i, a_{-i}).$$

Recursive structure

$$f_{i,t+1}(a_i) = \frac{1}{t+1} \sum_{t'=1}^{t+1} \mathbb{1}_{\{a_{i,t'}=a_i\}}$$

$$= \frac{1}{t+1} \sum_{t'=1}^{t} \mathbb{1}_{\{a_{j,t'}=a_j\}} + \frac{1}{t+1} \mathbb{1}_{\{a_{j,t+1}=a_i\}}$$

$$= \frac{t}{t+1} f_{i,t}(a_i) + \frac{1}{t+1} \mathbb{1}_{\{a_{j,t+1}=a_i\}}$$

$$= f_{i,t}(a_i) + \frac{1}{t+1} \left[\mathbb{1}_{\{a_{j,t+1}=a_i\}} - f_{i,t}(a_i) \right]$$

$$= f_{i,t}(a_i) + \lambda_i(t) \left[\mathbb{1}_{\{a_{j,t+1}=a_i\}} - f_{i,t}(a_i) \right]$$

where 1 is the indicator function [Brown 1951].

A reinforcement learning algorithm. $|A_i| < +\infty$, $\forall i \in \mathcal{K}, \forall n \in \{1, ..., |A_i|\},$

$$\pi_i^n(t+1) = \pi_i^n(t) + \lambda_i(t)u_i(t) \left[\mathbb{1}_{\{a_i(t)=a_i^n\}} - \pi_i^n(t) \right],$$

 $0 < \lambda_i(t) < 1.$

[Bush and Mosteller 1955][Sastry et al 1994].

- ► Required knowledge: individual utility <u>realizations</u>.
- ► Slow convergence.
- ► Steady state: NE/boundary points/limit cycle.

Convergence depends on: the updating rule + the associated game.

► For algorithms 1, 2, and 3, it is sufficient that the game be:

 \Box dominance solvable, or

 \Box potential, or

 \Box supermodular.

Exact potential games [Monderer and Shapley 1996]. $\exists \Phi, \forall i, \forall s, \forall s'_i, \forall s'_i, \forall s'_i, \forall s'_i, \forall s''_i, \forall s'''_i, \forall s''''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s''''_i, \forall s''''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s''''_i, \forall s''''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s''''_i, \forall s''''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s''''_i, \forall s''''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s'''_i, \forall s''''_i, \forall$

$$u_i(s) - u_i(s'_i, s_{-i}) = \Phi(s) - \Phi(s'_i, s_{-i}).$$

Characterization (special case). $S_i = I_i \subset \mathbb{R}$. A game is an exact PG iff:

$$\forall (i,j) \in \mathcal{K}^2, \ \frac{\partial^2 \left(u_i - u_j\right)}{\partial s_i \partial s_j} = 0.$$

Properties

- \blacktriangleright Convergence of important dynamics. \checkmark
- **Existence of a pure NE.** \checkmark

Examples. Team games, dummy games, self-motivated games, congestion games.

A simple example of potential game



A simple example of potential game [Perlaza et al 2009]

$$u_{i}(p_{1},...,p_{K}) = \sum_{s=1}^{S} \log \left(1 + \frac{g_{i,s}p_{i,s}}{\sigma^{2} + \sum_{j \neq i} g_{j,s}p_{j,s}} \right)$$
$$= \sum_{s=1}^{S} \log \left(\frac{\sigma^{2} + \sum_{j \neq i} g_{j,s}p_{j,s}}{\sigma^{2} + \sum_{j \neq i} g_{j,s}p_{j,s}} \right)$$
$$= \log \left(\sigma^{2} + \sum_{j} g_{j,s}p_{j,s} \right) - \log \left(\sigma^{2} + \sum_{j \neq i} g_{j,s}p_{j,s} \right)$$
$$= \Phi(p_{1},...,p_{K}) - \log \left(\sigma^{2} + \sum_{j \neq i} g_{j,s}p_{j,s} \right)$$

Another important special class of games: Supermodular games

Definition (supermodularity): S_i compact subset of \mathbb{R} , u_i upper semi-continuous in s, $\forall s_{-i} \ge s'_{-i}$, $u_i(s) - u_i(s_i, s'_{-i})$ is non-decreasing in s_i .

Characterization:

$$\forall i \neq j, \frac{\partial^2 u_i}{\partial s_i \partial s_j} \ge 0.$$

Properties

- \blacktriangleright Convergence of important dynamics. \checkmark
- \blacktriangleright Existence of a pure NE. \checkmark

Examples. Queueing problems [Yao 1995], power control problems [Saraydar et al 2002].

A simple example of supermodular game [Mochaourab & Jorswieck 2009]



 $u_1(\mu_1, \mu_2) = \log\left(1 + \rho g_{11,1}\mu_1\right) + \log\left(1 + \frac{\rho g_{11,2}\overline{\mu}_1}{1 + \rho g_{21,2}\mu_2}\right)$ $u_2(\mu_1, \mu_2) = \log\left(1 + \rho g_{22,1}\overline{\mu}_2\right) + \log\left(1 + \frac{\rho g_{22,2}\mu_2}{1 + \rho g_{12,2}\overline{\mu}_1}\right)$

Definition (regret) [Hart & Mas-Collel 2000]

$$\forall n, r_{k,a_{k,n}}(t+1) = \frac{1}{t} \sum_{t'=1}^{t} u_k(a_{k,n}, a_{-k}(t')) - u_k(a_k(t'), a_{-k}(t'))$$

Updating rule

$$\pi_{k,a_{k,n}}(t+1) = \frac{\left[r_{k,a_{k,n}}(t+1)\right]^{+}}{\sum_{n'=1}^{N_{k}} \left[r_{k,a_{k,n'}}(t+1)\right]^{+}}$$

Required knowledge: action profile

Convergence: unconditional convergence + intermediate speed

► Steady state: CCE

Remark: "pure NE \subseteq mixed NE \subseteq CE \subseteq CCE"

Definition

$$\forall k, \forall a'_k,$$
$$\sum_{a \in \mathcal{A}} q^{\text{CCE}}(a) u_k(a) \ge \sum_{a \in \mathcal{A}} q^{\text{CCE}}(a) u_k(a'_k, a_{-k})$$

Algorithms to reach a given solution concepts (strategic case)

- Asynchronous/synchronous best response dynamics, fictitious play, a type of reinforcement algorithm, regret matching,
- Boltzmann-Gibbs learning,
- coupled dynamics learning,
- trial-and-error learning,
- conditional no-regret learning,
- Bayesian learning,...

Game Theory: Fundamentals and Application to Wireless and Electricity Networks

4. Case study #2: Power consumption scheduling

- ► Static game formulation.
- ► A dynamic approach.

Application example. Continued



Modeling the problem as a static game

- ► Set of players : $\mathcal{I} = \{1, ..., I\}.$
- ► Set of actions : $s_i \in S_i = \{1, ..., T\}$

• Action profile : $s = (s_1, ..., s_I)$

► Total load :
$$\ell_t(s) = \ell_t^{exo} + \sum_i \ell_{i,t}^{EV}(s)$$

► Utilities :
$$u_i(s) = \sum_{t \in \{s_i, \dots, s_i + D_i - 1\}} f_t(\ell_1(s), \dots, \ell_t(s)) + g_i(s_i)$$

Nash equilibrium analysis (1/6)

Existence

Exact potentiality [Monderer Shapley 1996] $\exists \Phi, \forall i, \forall s, \forall s'_i$:

$$u_i(s) - u_i(s'_i, s_{-i}) = \Phi(s) - \Phi(s'_i, s_{-i})$$

Ordinal potentiality

$$u_i(s) - u_i(s'_i, s_{-i}) \ge 0 \Leftrightarrow \Phi(s) - \Phi(s'_i, s_{-i}) \ge 0$$

Result [Beaude et al TSG 2016]: OP available for memoryless utilities.

Nash equilibrium analysis (3/6)

Uniqueness No

Algorithm 1: The proposed distributed EV charging algorithm.

Initialize the round index as m = 0. Initialize the vector of charging start times as $s^{(0)}$. while $||s^{(m)} - s^{(m-1)}|| > \delta$ or $m \le M$ do Outer loop. Iterate on the round robin phase index: m = m + 1. Set i = 0. **Inner loop.** Iterate on the DM index: i = i + 1. Do: $s_i^{(m)} \in \arg\max_{s_i \in S_i} u_i(s_1^{(m)}, s_2^{(m)}, ..., s_i,$ $s_{i\pm 1}^{(m-1)}, \dots, s_{I}^{(m-1)}$ (11) where $s_i^{(m)}(i)$ stands for action of DM *i* in the round robin phase m. Stop when i = I and go to Outer loop. end



PoA: PoA \rightarrow 1 when $I \rightarrow \infty$ and under symmetry assumptions. Otherwise, losses may be non-negligible.

 \rightarrow Continuous actions

References: [Beaude et al Netgcoop 2012][Beaude et al TSG 2016][Paccagnan et al L-CSS 2018]
Why moving to a dynamical formulation?



Why moving to a dynamical formulation?



- ► Existence of an individual constraint on the state
- \blacktriangleright More efficiency: discrete actions \rightarrow continuous actions; exploit the dynamical structure
- ► Directly consider the global cost/utility function

$$\forall t, \qquad x_t \leq x_{\max}$$

$$\forall t \qquad x_t = ax_{t-1} + b_1 \times \left(\ell_t^{exo} + \sum_{i=1}^{I} v_{i,t} \right)^p + b_2 \times \left(\ell_{t-1}^{exo} + \sum_{i=1}^{I} v_{i,t-1} \right)^q + c_t$$

 $\forall t, \qquad x_t \leq x_{\max}$,

$$\begin{aligned} \forall (i,t), & 0 \leq v_{i,t} \leq V_{\max} \\ \forall t & x_t = ax_{t-1} + b_1 \times \left(\ell_t^{\exp} + \sum_{i=1}^I v_{i,t}\right)^p \\ & + b_2 \times \left(\ell_{t-1}^{\exp} + \sum_{i=1}^I v_{i,t-1}\right)^q + c_t \end{aligned}$$

 $\forall t, \qquad x_t \leq x_{\max}$,

$$\begin{aligned} \forall i, \qquad \sum_{t=1}^{T} v_{i,t} \geq C_i \\ \forall (i,t), \qquad 0 \leq v_{i,t} \leq V_{\max} \\ \forall t \qquad x_t = a x_{t-1} + b_1 \times \left(\ell_t^{\exp} + \sum_{i=1}^{I} v_{i,t} \right)^p \\ + b_2 \times \left(\ell_{t-1}^{\exp} + \sum_{i=1}^{I} v_{i,t-1} \right)^q + c_t \\ \forall t, \qquad x_t \leq x_{\max} , \end{aligned}$$

150

$$\begin{array}{ll} \text{minimize} & J(v,x) = \sum_{t=1}^{T} e^{\alpha x_t} + \gamma \left(\ell_t^{\text{exo}} + \sum_{i=1}^{I} v_{i,t} \right) \text{ s.t.} : \\ \forall i, & \sum_{t=1}^{T} v_{i,t} \geq C_i \\ \forall (i,t), & 0 \leq v_{i,t} \leq V_{\text{max}} \\ \forall t & x_t = a x_{t-1} + b_1 \times \left(\ell_t^{\text{exo}} + \sum_{i=1}^{I} v_{i,t} \right)^p \\ & + b_2 \times \left(\ell_{t-1}^{\text{exo}} + \sum_{i=1}^{I} v_{i,t-1} \right)^q + c_t \\ \forall t, & x_t \leq x_{\text{max}} , \end{array}$$

151

- **>** Substitution technique for x_t
- ► Operate in a convex regime (e.g., $ab_1 + b_2 \ge 0$)
- ► Apply the best response dynamics with $v_i = (v_{i,1}, ..., v_{i,T})$

[Beaude et al ECC 2015]

- ► Noisy forecast: $\tilde{\ell}_t^{\text{exo}} = \ell_t^{\text{exo}} + z_t$
- ▶ Randomness in the state evolution: $\tilde{c}_t = c_t + z'_t$
- Discretization + apply the best response dynamics with dynamical programming

[Gonzalez et al Gretsi 2017][Gonzalez et al TSG 2018]

Illustration 1



Illustration 2



Illustration 3



Game Theory: Fundamentals and Application to Wireless and Electricity Networks

4. Coalitional form games

Moving from strategic-form games to coalition form games

- Cooperation is sought/allowed
- Explicit communication is allowed
- ► Beyond NBS

The bankruptcy problem (Talmud's version)



Physical interpretation of the (game-theoretic) solution

[Aumann and Maschler 1985].



Coalition games can be a very powerful tool.

Two important issues in coalition games:

utility allocation/division;

► coalition formation.

Classification of coalition-form games



Coalition form games with characteristic functions

Definition. Game \equiv pair:

$$\mathcal{G} = (\mathcal{K}, v) \,.$$

Notation (power set). Ex: if $\mathcal{K} = \{1, 2\}, \ 2^{\mathcal{K}} = \{\emptyset, \{1\}, \{2\}, \{1, 2\}\}.$

Transferable utility (TU) games:

$$\begin{array}{rccc} v : & 2^{\mathcal{K}} & \to & \mathbb{R} \\ & \mathcal{C} & \mapsto & v(\mathcal{C}) \end{array}$$

Non transferable utility (NTU) games:

$$v: 2^{\mathcal{K}} \to \mathbb{R}^{\mathcal{K}}$$
$$\mathcal{C} \mapsto v(\mathcal{C}) = \{(v_1(\mathcal{C}), ..., v_K(\mathcal{C}))\}$$

Ice-cream game example (TU game) \rightarrow investors...



•
$$v(\emptyset) = v(\{C\}) = v(\{M\}) = v(\{T\}) = 0$$

• $v({C, M}) = 500, v({C, T}) = 500, v({M, T}) = 0$

Utility division: $x = (x_C, x_M, x_T)$.

- x = (200, 200, 350) not stable ($v(\{C, M\}) > x_C + x_M$).
- x' = (250, 250, 250) stable.
- x'' = (750, 0, 0) stable.

Notion of core (TU superadditive games):

$$\operatorname{core}(\mathcal{G}) = \left\{ x \in \mathbb{R}^{K} : \sum_{i \in \mathcal{K}} x_{i} = v(\mathcal{K}), \ \forall \mathcal{C} \subseteq \mathcal{K}, \sum_{i \in \mathcal{C}} x_{i} \ge v(\mathcal{C}) \right\}$$

Ice-cream game core

$$\begin{cases} x_1 + x_2 + x_3 &= ??? \\ x_1 &\geq ??? \\ x_2 &\geq ??? \\ x_3 &\geq ??? \\ x_1 + x_2 &\geq ??? \\ x_1 + x_3 &\geq ??? \\ x_2 + x_3 &\geq ??? \\ x_1 + x_2 + x_3 &\geq ??? \end{cases}$$

Ice-cream game core

Theorem (Bondareva-Shapley) Not treated here. See e.g., [Bacci et al 2016].

Definition (convex TU game)

 $\forall \mathcal{C}_1, \mathcal{C}_2 \subseteq \mathcal{K}, \ v(\mathcal{C}_1) + v(\mathcal{C}_2) \leq v(\mathcal{C}_1 \cup \mathcal{C}_2) + v(\mathcal{C}_1 \cap \mathcal{C}_2)$

Theorem Convex TU game \Rightarrow non-empty core.

The nucleolus

Core

$$\operatorname{core}(\mathcal{G}) = \Big\{ x \in \mathbb{R}^K : \sum_{i \in \mathcal{K}} x_i = v(\mathcal{K}), \, \forall \mathcal{C} \subseteq \mathcal{K}, v(\mathcal{C}) - \sum_{i \in \mathcal{C}} x_i \leq 0 \Big\}.$$

Excess:
$$e(x) = (e(\mathcal{C}_1, x), ..., e(\mathcal{C}_{2^K}, x))$$
 (with $e(\mathcal{C}_1, x) \ge e(\mathcal{C}_2, x) \ge ...$).

Nucleolus (relative to $\mathcal{X} \subseteq \mathbb{R}^K$)

nucleolus(
$$\mathcal{G}; \mathcal{X}$$
) = $\left\{ x \in \mathcal{X} : e(x) \preceq_L e(x'), \forall x' \in \mathcal{X} \right\}$.

Motivation Stability \rightarrow fairness

Definition Utility division:

$$x_i = \sum_{\mathcal{C} \subseteq \mathcal{K} \setminus \{i\}} \frac{|\mathcal{C}|!(|\mathcal{K}| - |\mathcal{C}| - 1)!}{|\mathcal{K}|!} \left[v(\mathcal{C} \cup \{i\}) - v(\mathcal{C}) \right].$$

Axiomatic characterization

► Efficiency:
$$\sum_{i \in \mathcal{K}} x_i = v(\mathcal{K})$$

► Additivity: $x_i(\mathcal{G}_1 \oplus \mathcal{G}_2) = x_i(\mathcal{G}_1) + x_i(\mathcal{G}_2)$ ($\oplus \equiv v = v_1 + v_2$)

► **Dummy:** $\forall C', v(C') = v(C' \cup \{i\})$ (*C* does not contain *i*)

► Symmetry: $\forall C'', v(C'' \cup \{i\}) = v(C'' \cup \{j\})$ (*C* does neither contain *i* nor *j*)

▶ **Players:** secondary transmitters $\mathcal{K} = \{1, ..., K\}$.

Characteristic function:

$$v(\mathcal{C}) = 1 - P_{\mathrm{m}}(\mathcal{C}) - J(P_{\mathrm{f}}(\mathcal{C}))$$

with

$$J(P_{\rm f}(\mathcal{C})) = \begin{vmatrix} -q^2 \log \left[1 - \left(\frac{P_{\rm f}(\mathcal{C})}{q}\right)^2 \right] & \text{if } 0 \le P_{\rm f}(\mathcal{C}) < q \\ +\infty & \text{if } q \le P_{\rm f}(\mathcal{C}) \le 1 \end{vmatrix}$$

Utility division. Not relevant.

Coalition formation. Merge and split coalitions by performing Pareto comparisons.

Results. Converging algorithm. Distributed solution: implementable, good performance in terms of miss and false alarm probabilities [Saad et al 2011].

- core, nucleolus,
- $\epsilon-$ core,
- least core,
- kernel,
- bargaining set,
- Shapley value, Harsanyi value, Banzhaf index,...

Game Theory: Fundamentals and Application to Wireless and Electricity Networks

5. Extensive form games

Definition: A standard extensive form game is a 6-uplet

$$\mathcal{G} = (\mathcal{K}, \mathcal{V}, v_{\text{root}}, \pi, \{\mathcal{V}_i\}_{i \in \mathcal{K}}, \{u_i\}_{i \in \mathcal{K}})$$

where:

- ▶ $\mathcal{K} = \{1, ..., K\}$ is the set of players;
- \blacktriangleright ($\mathcal{V}, v_{\text{root}}, \pi$) is a tree;
- ► $\{\mathcal{V}_i\}_{i \in \mathcal{K}}$ is a partition of \mathcal{V} .

Remark: $\forall v \in \mathcal{V}, \exists n \geq 1, \pi^{(n)} = \pi \circ ... \circ \pi = v_{\text{root}}.$

Representing the prisoner's dilemma under extensive form



Definition: It is a 9-uplet

$$\mathcal{G} = \left(\mathcal{K}, \mathcal{V}, v_{\text{root}}, \pi, \mathcal{V}_0, \{q_0^j\}_{j \in \mathcal{V}_0}, \{\mathcal{V}_i\}_{i \in \mathcal{K}}, \{W_i^k\}_{k \in \{1, \dots, k_i\}}, \{u_i\}_{i \in \mathcal{K}}\right)$$

where:

▶ player 0 is nature;

► $\forall j \in \mathcal{V}_0 q_j^0$ is the transition probability used by player 0 to choose a successor to j;

► W_i^k corresponds to the partition of V_i which defines the information structure for *i*.

Remark: Games with perfect information $W_i^k = \{w_i^k\}$.
- Extensive form **more complete** than strategic form.
- Extensive form usually **less convenient** for mathematical analysis.
- Continuous/discrete action sets.
- Extensive form sometimes more intuitive.
- The tree structure of the extensive form can be useful for computer-based analyses.

Transforming the prisoner's dilemma into a dynamic game



Game Theory: Fundamentals and Application to Wireless and Electricity Networks

6. Conclusion

► Direct game theory – mechanism design.

► 3 dominant mathematical representations: strategic form, extensive form, coalition form.

► Focus on the Nash equilibrium.

- ► Static games dynamic games.
- ► Relationship between static games and learning.

► Tradeoff between efficiency – weak information assumption.

Bridge the gap between learning and dynamic games.

Dynamic games with arbitrary observation graphs.

► Mechanism design.

Mechanism design, Nobel prizes,...

























Game Theory: Fundamentals and Application to Wireless and Electricity Networks

7. References

Quoted references (1)

Most quoted references can be found in:

Lasaulce and Tembine 2011: Lasaulce, S., & Tembine, H. (2011). Game theory and learning for wireless networks: fundamentals and applications. Academic Press.



Quoted references (2)

Knödel 1969: Knödel, W. (1969). Graphentheoretische Methoden und ihre Anwendungen. Springer-Verlag.

New York Times 1990: Kolata G. What if they closed 42d Street and nobody noticed? (1990, December). New York Times 1990-12-25.

Cohen and Kelly 1990: Cohen, J. E., & Kelly, F. P. (1990). A paradox of congestion in a queuing network. Journal of Applied Probability, 730-734.

Perlaza et al 2009: Perlaza, S. M., Belmega, E. V., Lasaulce, S., & Debbah, M. (2009, October). On the base station selection and base station sharing in self-configuring networks. In Proceedings of the Fourth International ICST Conference on Performance Evaluation Methodologies and Tools (p. 71).

Björnson et al 2015: Bjornson, E., Jorswieck, E., Debbah, M., & Ottersten, B. (2014). Multiobjective Signal Processing Optimization: The way to balance conflicting metrics in 5G systems. Signal Processing Magazine, IEEE, 31(6), 14-23.

Bacci et al 2015: Bacci, G., Lasaulce, S., Saad, W., & Sanguinetti, L. (2015). Game Theory for Signal Processing in Networks. arXiv preprint arXiv:1506.00982.

Quoted references (3)

Cournot 1838: Cournot, A. A. (1838). Recherches sur les principes mathmatiques de la thorie des richesses par Augustin Cournot. chez L. Hachette.

Nash 1950: Nash, J. F. (1950). Equilibrium points in n-person games. Proc. Nat. Acad. Sci. USA, 36(1), 48-49.

Kuhn 1953: Kuhn, H. W. (1953). Extensive games and the problem of information. Contributions to the Theory of Games, 2(28), 193-216.

Glicksberg 1952: Glicksberg, I. L. (1952). A further generalization of the Kakutani fixed point theorem, with application to Nash equilibrium points. Proceedings of the American Mathematical Society, 3(1), 170-174.

Debreu 1952: Debreu, G. (1952). A social equilibrium existence theorem. Proceedings of the National Academy of Sciences of the United States of America, 38(10), 886.

Fan 1952: Fan, K. (1952). Fixed-point and minimax theorems in locally convex topological linear spaces. Proceedings of the National Academy of Sciences of the United States of America, 38(2), 121.

Quoted references (4)

Rosen 1965: Rosen, J. B. (1965). Existence and uniqueness of equilibrium points for concave n-person games. Econometrica: Journal of the Econometric Society, 520-534.

Yates 1995: Yates, R. D. (1995). A framework for uplink power control in cellular radio systems. Selected Areas in Communications, IEEE Journal on, 13(7), 1341-1347.

Perlaza & Lasaulce 2014: Perlaza, S. M., & Lasaulce, S. (2014). Game-Theoretic Solution Concepts and Learning Algorithms. Mechanisms and Games for Dynamic Spectrum Allocation, 185-221.

Yates 1995: Yates, R. D. (1995). A framework for uplink power control in cellular radio systems. Selected Areas in Communications, IEEE Journal on, 13(7), 1341-1347.

Papadimitriou 2001: Papadimitriou, C. (2001, July). Algorithms, games, and the internet. In Proceedings of the thirty-third annual ACM symposium on Theory of computing (pp. 749-753). ACM.

Quoted references (5)

Anshelevich 2014: Anshelevich, E., Dasgupta, A., Kleinberg, J., Tardos, E., Wexler, T., & Roughgarden, T. (2004). The price of stability for network design with selfish agents. In IEEE Symposium on Foundations of Computer Science (pp. 295-304).

Correa et al 2005: Correa, J. R., Schulz, A. S., & Stier-Moses, N. E. (2005). On the inefficiency of equilibria in congestion games. In Integer Programming and Combinatorial Optimization (pp. 167-181). Springer Berlin Heidelberg.

Larsson & Jorswieck 2008: Larsson, E. G., & Jorswieck, E. (2008). Competition versus cooperation on the MISO interference channel. Selected Areas in Communications, IEEE Journal on, 26(7), 1059-1069.

Yu et al 2002: Yu, W., Ginis, G., & Cioffi, J. M. (2002). Distributed multiuser power control for digital subscriber lines. Selected Areas in Communications, IEEE Journal on, 20(5), 1105-1115.

Quoted references (6)

Brown 1951: Brown, G. W. (1951). Iterative solution of games by fictitious play. Activity analysis of production and allocation, 13(1), 374-376.

Bush & Mosteller 1955: Bush, R. R., & Mosteller, F. (1955). Stochastic models for learning.

Sastry et al 1994: Sastry, P. S., Phansalkar, V. V., & Thathachar, M. (1994). Decentralized learning of Nash equilibria in multi-person stochastic games with incomplete information. Systems, Man and Cybernetics, IEEE Transactions on, 24(5), 769-777.

Monderer & Shapley 1996: Monderer, D., & Shapley, L. S. (1996). Potential games. Games and economic behavior, 14(1), 124-143.

Yao 1995: Yao, D. D. (1995). S-modular games, with queueing applications. Queueing Systems, 21(3-4), 449-475.

Saraydar 2002: Saraydar, C. U., Mandayam, N. B., & Goodman, D. J. (2002). Efficient power control via pricing in wireless data networks. Communications, IEEE Transactions on, 50(2), 291-303.

Quoted references (7)

Hart and Mas-Collel & 2000: Hart, S., & Mas-Collel, A. N. D. R. E. U. (2000). A simple adaptive procedure leading to correlated equilibrium. Econometrica, 68(5), 1127-1150.

Mochaourab & Jorswieck 2009: Mochaourab, R., & Jorswieck, E. (2009, October). Resource allocation in protected and shared bands: uniqueness and efficiency of Nash equilibria. In Proceedings of the Fourth International ICST Conference on Performance Evaluation Methodologies and Tools (p. 68).

Renault & Tomala 1998: Renault, J., & Tomala, T. (1998). Repeated proximity games. International Journal of Game Theory, 27(4), 539-559.

Goodman & Mandayam 2000: Goodman, D., & Mandayam, N. (2000). Power control for wireless data. Personal Communications, IEEE, 7(2), 48-54.

Le Treust & Lasaulce 2010: Le Treust, M., & Lasaulce, S. (2010). A repeated game formulation of energy-efficient decentralized power control. Wireless Communications, IEEE Transactions on, 9(9), 2860-2869.

Mériaux et al 2011: Mériaux, F., Le Treust, M., Lasaulce, S., & Kieffer, M. (2011, July). A stochastic game formulation of energy-efficient power control: Equilibrium utilities and practical strategies. In Digital Signal Processing (DSP), 2011 17th International Conference on (pp. 1-6). IEEE.

Quoted references (8)

Larrousse and Lasaulce 2015: B. Larrousse and S. Lasaulce, "Coded Power Control: Performance Analysis", IEEE Intl. Symposium on Information Theory (ISIT), Istanbul, Turkey, Jul. 2013.

Larrousse et al 2015: Larrousse, B., Lasaulce, S. & Wigger, M. (2015, May). Coordination in State-Dependent Distributed Networks. In Information Theory Workshop (ITW). (pp. 1-5). IEEE.

Larrousse et al 2018: B. Larrousse, S. Lasaulce, and M. Bloch, Coordination in distributed networks via coded actions with application to power control, IEEE Transactions on Information Theory, Vol. 64, No. 5, pp. 3633-3654, 2018.

Gossner et al 2006: O. Gossner, P. Hernández, and A. Neyman, "Optimal use of communication resources", Econometrica, Vol.74, No.6, pp. 1603-1636, 2006.

Agrawal et al 2015: Agrawal, A., Lasaulce, S. & Visoz, R. (2015, November). A framework for decentralized power control with partial channel state information. In Communications and Networking (ComNet), 2015 4th International Conference on (pp. 1-6). IEEE.

Aumann & Maschler 1985: Aumann, R. J., & Maschler, M. (1985). Game theoretic analysis of a bankruptcy problem from the Talmud. Journal of economic Theory, 36(2), 195-213.

Quoted references (9)

[Aumann and Maschler 1985] Aumann, R.J., Maschler, M., 1985. Game theoretic analysis of a bankruptcy problem from the Talmud. J. Econ. Theory, 36, 195213.

Saad et al 2011: Saad, W., Han, Z., Başar, T., Debbah, M., & Hjørungnes, A. (2011). Coalition formation games for collaborative spectrum sensing. Vehicular Technology, IEEE Transactions on, 60(1), 276-297.

[Gonzalez et al Gretsi 2017] M. Gonzalez, O. Beaude, P. Bouyer, S. Lasaulce, and N. Markey, "Stratgies dordonnancement de consommation dnergie en prsence dinformation imparfaite de prvision", Gretsi conference, Juan-les-Pins, France, Sep. 2017.

[Gonzalez et al TSG 2018] M. Gonzalez, O. Beaude, P. Bouyer, S. Lasaulce, and N. Markey, "Dynamic load scheduling under uncertainty", IEEE Trans. on Smart Grid, submitted.

[Bacci et al 2016] G. Bacci, S. Lasaulce, W. Saad, and L. Sanguinetti, "Game theory and its applications in signal processing", IEEE Signal Processing Magazine, Vol. 33, No. 1, pp. 94–119, Jan. 2016.

G. Bacci, S. Lasaulce, W. Saad, and L. Sanguinetti, "Game theory and its applications in signal processing", IEEE Signal Processing Magazine, Vol. 33, No. 1, pp. 94–119, Jan. 2016.

Quoted references (10)

[Beaude et al ECC 2015] O. Beaude, S. Lasaulce, M. Hennebel, and J. Daafouz, "Minimizing the impact of EV charging on the distribution network", European Conference on Control (ECC), Linz, Austria, July 2015.

[Beaude et al TSG 2016] O. Beaude, S. Lasaulce, M. Hennebel, and I. Mohand-Kaci, "Reducing the impact of distributed EV charging on distribution network operating costs", IEEE Transactions on Smart Grid, Vol. 7, No.6, pp. 2666–2679, 2016.

[Paccagnan et al L-CSS 2018] D. Paccagnan, F. Parise, and J. Lygeros, "On the Efficiency of Nash Equilibria in Aggregative Charging Games", IEEE L-CSS, 2018.

[Perlaza et al 2013] S. M. Perlaza, S. Lasaulce, and M. Debbah, "Equilibria of Channel Selection Games in Parallel Multiple Access Channels", EURASIP Journal on Wireless Communications and Networking (JWCN), Jan. 2013.

[Sastry et al 1994] Sastry, P.S., Phansalkar, V.V., Thatchar, M.A.L., 1994. Decentralized learning of Nash equilibria in multi-person stochastic games with incomplete information. IEEE Trans. Syst. Man Cybern. 24, 769777.