

Seeking Nash Equilibria under Nonconvex Coupling Constraints

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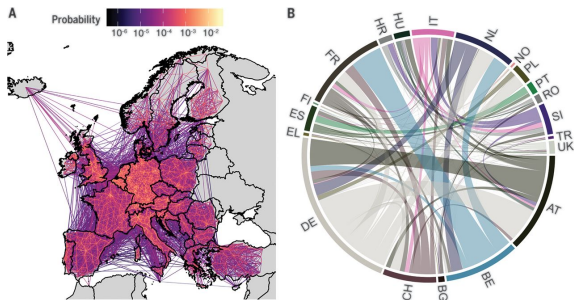
Introduction

Why do we need Game Theory?

COVID-19 pandemic: independent decision-makers (EU commission, governments, municipalities, institutions, citizens)

Uncoordinated reactions!

The selfish strategy of a decision-maker impacted all the others



From Ruktanonchai et al., *Assessing the impact of coordinated COVID-19 exit strategies*, Science, 2020

- ▷ **Players or decision-makers:** $i \in \mathcal{N} := \{1, \dots, N\} \subseteq \mathbb{N}$
- ▷ **Actions or strategies:** decision variable $\mathbf{x}_i \in \Omega_i \subseteq \mathbb{R}^n$

$$\mathbf{x} = \begin{pmatrix} \mathbf{x}_1 \\ \vdots \\ \mathbf{x}_N \end{pmatrix} \in \mathbb{R}^{(N)n}, \quad \forall i \in \mathcal{N} : \mathbf{x}_{-i} = \begin{pmatrix} \mathbf{x}_1 \\ \vdots \\ \mathbf{x}_{i-1} \\ \mathbf{x}_{i+1} \\ \vdots \\ \mathbf{x}_N \end{pmatrix} \in \mathbb{R}^{(N-1)n}$$

- ▷ **Outcomes or payoff functions:** $f_i(\mathbf{x}_i, \mathbf{x}_{-i}) : \mathbb{R}^n \times \mathbb{R}^{(N-1)n} \rightarrow \mathbb{R}$
 - ▷ **Game:** the "outcome" is dependent also on \mathbf{x}_{-i} , i.e., the strategies of the other players

Several alternative solutions for the game: Nash equilibrium (NE)

Definition: Nash equilibrium

$$f_i(\mathbf{x}_i^*, \mathbf{x}_{-i}^*) \leq \inf \{f_i(\mathbf{x}_i, \mathbf{x}_{-i}^*) \mid \mathbf{x}_i \in \Omega_i\}, \quad \forall i \in \mathcal{N}.$$

Nash equilibrium problem (NEP):

$$\forall i \in \mathcal{N} : \begin{cases} \min_{\mathbf{x}_i} & f_i(\mathbf{x}_i, \mathbf{x}_{-i}) \\ \text{subject to} & \mathbf{x}_i \in \Omega_i \end{cases}$$

From a control-theoretic perspective, the objective is to develop a coordination mechanism for updating players' strategies to a NE.



Nash, Equilibrium points in N-person games, PNAS, 1950

Generalized Nash equilibrium problem

The strategies that a player can choose may be limited by the strategies of other players \mathbf{x}_{-i} , i.e., exists a coupled feasible set

$$\mathcal{X} = \Omega \cap \{\mathbf{x} \in \mathbb{R}^{Nn} \mid g(\mathbf{x}) \leq \mathbf{0}_M\}$$

hence $\mathbf{x}_i \in \mathcal{X}_i(\mathbf{x}_{-i})$

Definition: Generalized Nash equilibrium

$$f_i(\mathbf{x}_i^*, \mathbf{x}_{-i}^*) \leq \inf \{f_i(\mathbf{x}_i, \mathbf{x}_{-i}^*) \mid \mathbf{x}_i \in \mathcal{X}_i(\mathbf{x}_{-i}^*)\}, \forall i \in \mathcal{N}.$$

Generalized Nash equilibrium problem (GNEP):

$$\forall i \in \mathcal{N} : \begin{cases} \min_{\mathbf{x}_i} & f_i(\mathbf{x}_i, \mathbf{x}_{-i}) \\ \text{s.t.} & \mathbf{x}_i \in \mathcal{X}_i(\mathbf{x}_{-i}). \end{cases}$$



Debreu, A social equilibrium existence theorem, PNAS, 1952

From an optimization point of view: a point \mathbf{x}_i^* is said to be an optimal solution for a player if:

$$f(\mathbf{x}_i^*) \leq f(\mathbf{x}_i), \quad \forall \mathbf{x}_i \in \mathcal{X}_i(\mathbf{x}_{-i}^*)$$

when a point is an optimal solution? \Rightarrow **optimality conditions**

Definition: Minimum principle for single valued functions

A feasible point \mathbf{x}_i^* is an optimal solution if and only if

$$(\mathbf{y} - \mathbf{x}_i^*)^\top \nabla f(\mathbf{x}_i^*) \geq 0, \quad \forall \mathbf{y} \in \mathcal{X}_i(\mathbf{x}_{-i}^*)$$

equivalent to the Karush-Kuhn-Tucker (KKT) conditions when the set is defined by inequalities and equalities

How to prove the existence of a generalized Nash equilibrium?

...is quite difficult, we need some technical assumptions...

- ▷ **Continuity** of $f_i(\mathbf{x}_i, \mathbf{x}_{-i})$ and **compactness** of Ω_i
 - ▷ can be relaxed with further assumptions
- ▷ **Convexity** of $f_i(\mathbf{x}_i, \mathbf{x}_{-i})$, Ω_i and \mathcal{X} required for fixed-point theorems (Brouwer, Kakutani, Banach-Picard, Caristi)

Jointly Convex Nash Equilibrium Problem

- ▷ Existence and uniqueness
- ▷ Convergence guarantee

sets $\mathcal{X}_i(\mathbf{x}_{-i})$ defined by a system of inequalities, componentwise convex with respect to all variables.



Facchinei, Fischer, Piccialli, On generalized Nash games and variational inequalities, ORL, 2007

Monotone Operators and Variational Inequality Theory

Definition: Variational inequality problem

The *variational inequality problem* $VI(\mathcal{X}, F(\mathbf{x}))$ consists in finding $\mathbf{y} \in \mathcal{X}$ such that:

$$\inf_{\mathbf{x} \in \mathcal{X}} (\mathbf{x} - \mathbf{y})^\top F(\mathbf{y}) \geq 0.$$

$\mathcal{X} \subseteq \mathbb{R}^n$ closed and convex subset and $F : \mathcal{X} \rightarrow \mathbb{R}^n$ continuous map.

Definition: Quasi-variational inequality problem

The *quasi-variational inequality problem* $QVI(\mathcal{X}(\mathbf{x}), F(\mathbf{x}))$ consists in finding $\mathbf{y} \in \mathcal{X}(\mathbf{x})$ such that:

$$\inf_{\mathbf{x} \in \mathcal{X}(\mathbf{x})} (\mathbf{x} - \mathbf{y})^\top F(\mathbf{y}) \geq 0.$$

$\mathcal{X}(\mathbf{x})$ point-to-set mapping and $F : \mathcal{X} \rightarrow \mathbb{R}^n$ continuous map.

Solution set

- ▷ If F is monotone on \mathcal{X} , the solution set of VI, is closed and convex.

$$(F(\mathbf{x}) - F(\mathbf{y}))^\top (\mathbf{x} - \mathbf{y}) \geq 0, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n$$

- ▷ If F is strongly monotone on \mathcal{X} with a constant $\mu \in \mathbb{R}_{>0}$, the solution set of VI, admits a unique solution.

$$(F(\mathbf{x}) - F(\mathbf{y}))^\top (\mathbf{x} - \mathbf{y}) \geq \mu \|\mathbf{x} - \mathbf{y}\|^2, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n$$

QVIs are much harder to solve!

We can solve some classes of (G)NEP by finding a solution for the associated *(quasi-)variational inequality problem!*

VI associated with a NEP: $VI(\Omega, F(\mathbf{x}))$

$$F(\mathbf{x}) = \begin{bmatrix} \nabla_{\mathbf{x}_1} f_1(\mathbf{x}_1, \mathbf{x}_{-1}) \\ \vdots \\ \nabla_{\mathbf{x}_N} f_i(\mathbf{x}_N, \mathbf{x}_{-N}) \end{bmatrix}, \quad \Omega = \prod_{i=1}^N \Omega_i$$

Solve the NEP \Leftrightarrow Solve $VI(\Omega, F(\mathbf{x}))$

QVI associated with a GNEP: $QVI(\mathcal{X}, F(\mathbf{x}))$

$$\mathcal{X} = \Omega \cap \{\mathbf{x} \in \mathbb{R}^{Nn} \mid g(\mathbf{x}) \leq \mathbf{0}_M\} \quad \text{or} \quad \mathcal{X} = \prod_{i=1}^N \mathcal{X}_i(\mathbf{x}_{-i})$$

Solve the GNEP \Leftrightarrow Solve $QVI(\mathcal{X}, F(\mathbf{x}))$

Jointly convex case:

Solve the GNEP \Leftarrow Solve $VI(\mathcal{X}, F(\mathbf{x}))$

Characterization of the solution obtained by the VI

$$\begin{aligned}\forall i \in \mathcal{N} : \quad & \nabla_{\mathbf{x}_i} f_i(\mathbf{x}_i, \mathbf{x}_{-i}) + \nabla_{\mathbf{x}_i} \phi(\mathbf{x}_i, \mathbf{x}_{-i}) \boldsymbol{\lambda}_i = \mathbf{0}_{Nn} \\ & \boldsymbol{\lambda}_i \geq \mathbf{0}_{Nn} \perp \phi(\mathbf{x}_i, \mathbf{x}_{-i}) \leq \mathbf{0}_{Nn} \\ & (\mathbf{x} - \mathbf{x}^*)^\top F(\mathbf{x}^*) \geq 0, \quad \forall \mathbf{x} \in \mathcal{X}.\end{aligned}$$

KKT conditions of the corresponding VI($\mathcal{X}, F(\mathbf{x})$):

$$\begin{aligned}\text{VI} : \quad & \begin{pmatrix} \nabla_{\mathbf{x}_1} f_1(\mathbf{x}_1, \mathbf{x}_{-1}) \\ \vdots \\ \nabla_{\mathbf{x}_i} f_i(\mathbf{x}_i, \mathbf{x}_{-i}) \end{pmatrix} + \begin{pmatrix} \nabla_{\mathbf{x}_1} \phi(\mathbf{x}) \\ \vdots \\ \nabla_{\mathbf{x}_i} \phi(\mathbf{x}) \end{pmatrix} \boldsymbol{\lambda} = \mathbf{0} \\ & \boldsymbol{\lambda} \geq \mathbf{0} \perp \phi(\mathbf{x}) \leq 0\end{aligned}$$

Variational solutions: GNEP that are preserved passing to the VIP are exactly those for which all players have the same multipliers for the respective constraints: **"fair" behaviour**

There is a convexity for the mapping F ?

Monotonicity

- ▷ If f is convex $\Leftrightarrow F$ is monotone
- ▷ If f is strictly convex $\Leftrightarrow F$ is strictly monotone
- ▷ If f is strongly convex $\Leftrightarrow F$ is strongly monotone

Theorem: Existence of a GNE

Let us consider a jointly convex generalized game. A solution of the $VI(\mathcal{X}, F(x))$ **exists**, is **unique** and is a solution of the original GNEP.



Harker, Generalized Nash games and quasi-variational inequalities. EJOR, 1991.



Facchinei, Kanzow. Generalized Nash equilibrium problems. AOR, 2007.

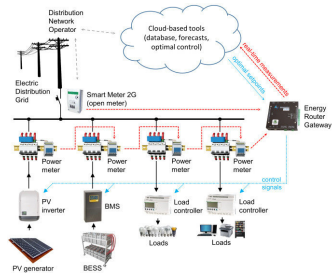
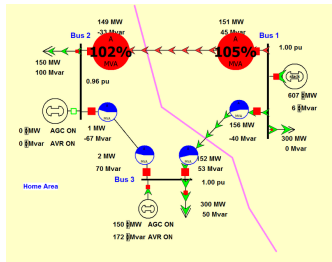
Motivating Examples

Noncooperative Optimal Power Flow in Power Grids

Nonconvex constraints: power flow equations

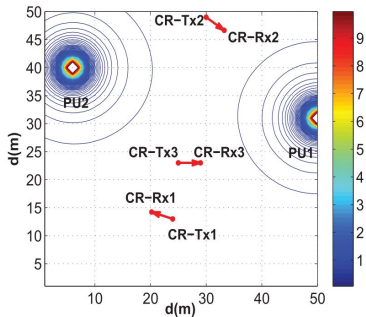
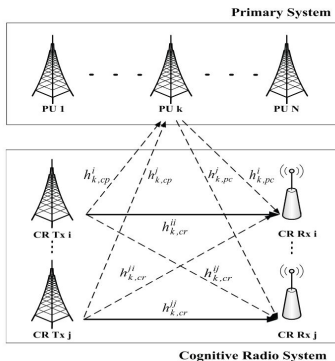
$$|V_b| \sum_{r \in \mathcal{B}} |V_r| |Y_{br}| \cos(\theta_{br} + \theta_r - \theta_b) = P_b$$

$$-|V_b| \sum_{r \in \mathcal{B}} |V_r| |Y_{br}| \sin(\theta_{br} + \theta_r - \theta_b) = Q_b$$



Power Allocation Games in Cognitive Radio Systems

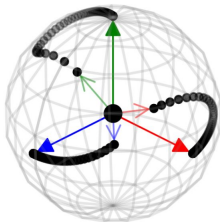
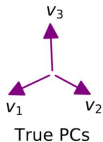
Nonconvex constraints: power allocation in Gaussian frequency-selective interference channels



Eigengame: PCA as a Nash equilibrium

Principal component analysis (PCA) as a competitive game in which each approximate eigenvector is controlled by a player whose goal is to maximize their own utility function

parallelizing the computation by transforming the problem in a noncooperative game



Clarke's Local Generalized Nash Equilibria

Due to the nature of several applications, the coupling feasible set \mathcal{X} may result to be nonconvex.

Assumption 1: cost functions

For each $i \in \mathcal{N}$ and for every \mathbf{x}_{-i} , $f_i(\cdot, \mathbf{x}_{-i})$ convex and continuously differentiable.

Assumption 2: coupling feasible set

For each $m \in \mathcal{M}$ and for every \mathbf{x}_{-i} , $g_m(\cdot, \mathbf{x}_{-i})$ continuously differentiable (possibly nonconvex). \mathcal{X} is nonempty and compact.

We analyze the particular case of nonconvex feasible sets... with nonconvex feasible sets.

Weaker equilibrium conditions:

- ▶ Clarke's local generalized Nash equilibrium (CL-GNE).

Clarke's tangent cone $\tilde{\mathcal{X}}(\mathbf{x}) := \mathbf{x} + T_{\text{cl}}(\mathcal{X}, \mathbf{x})$ at a point \mathbf{x} for the (nonconvex) set \mathcal{X} .

Definition: Clarke's local GNE

A CL-GNE is a collective strategy $\mathbf{x}^* \in \mathcal{X}$ such that for each $i \in \mathcal{N}$:

$$f_i(\mathbf{x}_i^*, \mathbf{x}_{-i}^*) \leq \inf \{f_i(\mathbf{y}, \mathbf{x}_{-i}^*) \mid \mathbf{y} \in \tilde{\mathcal{X}}_i(\mathbf{x}_{-i}^*)\}$$

- ▶ If \mathcal{X} is convex, the CL-GNE is equivalent to the GNE
- ▶ **Approximation of the set in a specific point.**
- ▶ Always convex even if \mathcal{X} is nonconvex

Clarke's tangent vector

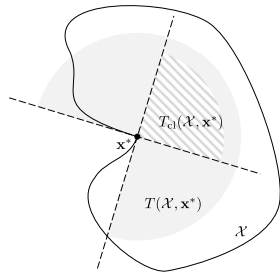
Nonempty subset \mathcal{X} of a Euclidean space E , and a point $\mathbf{x} \in \text{cl}(\mathcal{X})$.
 $(\mathbf{x}_n)_{n \in \mathbb{N}} \in \mathcal{X}$, $(t_n)_{n \in \mathbb{N}}$ $(h_n)_{n \in \mathbb{N}} \in E$ such that:

$$\mathbf{x} = \lim_{n \rightarrow \infty} \mathbf{x}_n, \quad \lim_{n \rightarrow \infty} t_n = 0, \quad \mathbf{d} = \lim_{n \rightarrow \infty} d_n, \quad \mathbf{x}_n + t_n d_n \in \mathcal{X}$$

$\mathbf{d} \in E$ is a *Clarke's tangent vector* to \mathcal{X} at \mathbf{x} .

Definition: Clarke's tangent cone

The set of all Clarke's tangent vectors to \mathcal{X} at \mathbf{x} is the *Clarke's tangent cone* $T_{\text{cl}}(\mathcal{X}, \mathbf{x})$.



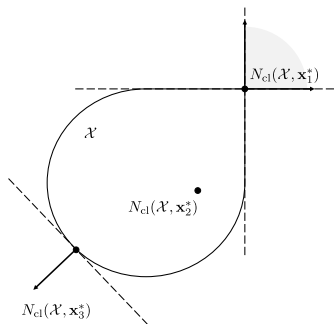
For each player $i \in \mathcal{N}$:

▷ optimality condition

$$-\nabla f_i(\mathbf{x}_i^*, \mathbf{x}_{-i}^*) \in N_{\text{cl}}(\mathcal{X}_i(\mathbf{x}_{-i}^*), \mathbf{x}_i^*)$$

▷ Karush-Kuhn-Tucker (KKT) conditions

$$\text{KKT}_i : \begin{cases} 0 \in \nabla_{\mathbf{x}_i} f_i(\mathbf{x}_i^*, \mathbf{x}_{-i}^*) + \nabla_{\mathbf{x}_i} g(\mathbf{x}_i^*, \mathbf{x}_{-i}^*) \boldsymbol{\lambda}_i \\ \mathbf{0}_M \leq \boldsymbol{\lambda}_i \perp g(\mathbf{x}_i^*, \mathbf{x}_{-i}^*) \leq \mathbf{0}_M \end{cases}$$



Theorem: Characterization

If $\mathbf{x}^* \in \mathcal{X}$ is a CL-GNE, we have that for each $i \in \mathcal{N}$:

(i) $-\nabla f_i(\mathbf{x}_i^*, \mathbf{x}_{-i}^*) \in N_{\text{cl}}(\mathcal{X}_i(\mathbf{x}_{-i}^*), \mathbf{x}_i^*)$;

(ii) there exists a vector $\boldsymbol{\lambda}_i \in \mathbb{R}_{\geq 0}^M$ satisfying the KKT conditions.

We define a *quasi-variational inequality problem* $QVI(\tilde{\mathcal{X}}(\mathbf{x}), F(\mathbf{x}))$

$$\inf_{\mathbf{y} \in \tilde{\mathcal{X}}(\mathbf{x}^*)} (\mathbf{y} - \mathbf{x}^*)^\top F(\mathbf{x}^*) \geq 0.$$

We define the *variational Clarke local generalized Nash equilibrium* (vCL-GNE) as a solution of the CL-GNEP that satisfies the QVIP.

Similarly to the relation between GNEP and VIP, **not all the solutions of the CL-GNEP are solutions of the QVIP**; viceversa, **a solution of the QVIP is a solution of the original CL-GNEP**.

- ▶ **Solve the GNEP** \Leftrightarrow **Solve** $QVI(\mathcal{X}, F(\mathbf{x}))$
- ▶ **Solve the GNEP** \Leftarrow **Solve** $VI(\mathcal{X}, F(\mathbf{x}))$ (Jointly convex case)
- ▶ **Solve the CL-GNEP** \Leftarrow **Solve** $QVI(\tilde{\mathcal{X}}(\mathbf{x}), F(\mathbf{x}))$

$\tilde{\mathcal{X}}(\mathbf{x})$ is convex: only variational solutions!

At a vCL-GNE, we have that in a local subset of \mathcal{X} each agent cannot unilaterally maximize their own function while keeping the strategies of the other agents fixed (**locally fair equilibrium point**).

Theorem: Characterization

- (i) Let \mathbf{x}^* be a solution of the CL-GNEP, where the KKT conditions for all players hold with the same Lagrangian multipliers $\boldsymbol{\lambda} = \boldsymbol{\lambda}_i, \forall i \in \mathcal{N}$. Then, \mathbf{x}^* is a solution of the QVI and thus it is a vCL-GNE.
- (ii) Viceversa, let \mathbf{x}^* be a solution of the QVI and thus be a vCL-GNE. Then, \mathbf{x}^* is a solution of the CL-GNEP at which the KKT conditions hold with the same Lagrangian multipliers, $\boldsymbol{\lambda} = \boldsymbol{\lambda}_i, \forall i \in \mathcal{N}$.

Example 1

Two players, with strategies $\mathbf{x}_1 \in \mathbb{R}$ and $\mathbf{x}_2 \in \mathbb{R}$

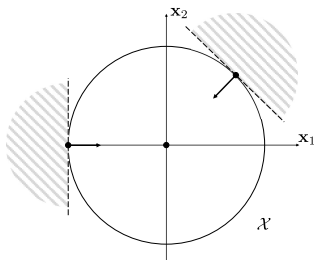
$$\forall i \in \{1, 2\} : \begin{cases} \min_{-2 \leq \mathbf{x}_i \leq 2} & f_i(\mathbf{x}_i) = \mathbf{x}_i^2 \\ \text{s.t.} & \mathbf{x}_1^2 + \mathbf{x}_2^2 \geq 1. \end{cases}$$

Cost functions decoupled and strictly convex.

The game has infinite CL-GNE

$$\begin{aligned} 2\mathbf{x}_i - \lambda_i 2\mathbf{x}_i &= 0 \\ \lambda_i \geq 0 \perp \mathbf{x}_1^2 + \mathbf{x}_2^2 &\geq 1 \quad \forall i = 1, 2. \end{aligned}$$

All points on the unitary circumference $\lambda_1 = \lambda_2 = 1$; hence, all these points are also vCL-GNE.



Example 2

Modified version of Example 1.

$$\forall i \in \{1, 2\} : \begin{cases} \min & f_i(\mathbf{x}_i) = \mathbf{x}_i^2 \\ \text{s.t.} & (\mathbf{x}_1 - \frac{1}{4})^2 + (\mathbf{x}_2 - \frac{1}{4})^2 \geq 1. \end{cases}$$

$\forall i = 1, 2$

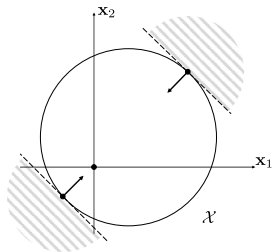
$$2\mathbf{x}_i - \lambda_i(2\mathbf{x}_i - \frac{1}{2}) = 0$$

$$\lambda_i \geq 0 \perp 1 - (\mathbf{x}_1 - \frac{1}{4})^2 - (\mathbf{x}_2 - \frac{1}{4})^2 \leq 0$$

Only two vCL-GNE

$$\triangleright \mathbf{x}_1 = \mathbf{x}_2 = \frac{1}{4} + \frac{\sqrt{2}}{2}$$

$$\triangleright \mathbf{x}_1 = \mathbf{x}_2 = \frac{1}{4} - \frac{\sqrt{2}}{2}$$



Existence and Uniqueness

- ▷ Projection operator

$$\text{proj}_{\mathcal{X}}(\mathbf{y}) := \operatorname{argmin}_{\mathbf{x} \in \mathcal{X}} \|\mathbf{x} - \mathbf{y}\|$$

- ▷ Distance operator between a point and a set

$$\text{dist}(\mathbf{y}, \mathcal{X}) := \min_{\mathbf{x} \in \mathcal{X}} \|\mathbf{x} - \mathbf{y}\|$$

$F(\cdot) : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is Lipschitz continuous with a constant $\ell \in \mathbb{R}_{>0}$ if

$$\|F(\mathbf{x}) - F(\mathbf{y})\| \leq \ell \|\mathbf{x} - \mathbf{y}\|, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n$$

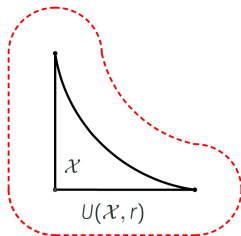
- ▷ for $\ell = 1$, we say F is **nonexpansive**
- ▷ for $\ell < 1$, we say F is a **contraction** (with contraction factor ℓ)

If \mathcal{X} is convex $\text{proj}_{\mathcal{X}}(\mathbf{y})$ is a **contraction**: existence and convergence proofs based on projected gradient $\mathbf{x}^* \in \text{fix}(\text{proj}_{\mathcal{X}}(\text{Id} - \gamma F(\cdot)))$

The projection onto **nonconvex** sets is **not a nonexpansive operator**.

- ▶ Can we ensure that $\text{proj}_{\mathcal{X}}(\mathbf{y})$ is a contraction when \mathcal{X} is nonconvex?
- ▶ Can we still work with a nonexpansive operator (or expansive with a certain degree)?

For any set \mathcal{X} and $r > 0$, one sets
 $U(\mathcal{X}, r) := \{\mathbf{r} \in \mathbb{R}^n \mid \text{dist}(\mathbf{y}, \mathcal{X}) < r\}$



Definition: Proximally Smooth Sets

A set $\mathcal{X} \subseteq \mathbb{R}^n$ is said to be proximally smooth if there exists $r > 0$ such that the distance function $\text{dist}(\cdot, \mathcal{X})$ is continuously differentiable on the r -enlargement $U(\mathcal{X}, r) := \mathcal{X} + r\mathbb{B}$.



Federer, Curvature measures. TAMS, 1959.

Proximal smoothness properties

Let $\mathcal{X} \subseteq \mathbb{R}^n$ be a nonempty closed set. If \mathcal{X} is r -proximally smooth, then the following properties hold for any $r' \in (0, r)$:

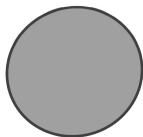
- (i) $\text{proj}_{\mathcal{X}}(\mathbf{x}) \neq \emptyset$ for all $\mathbf{x} \in U(\mathcal{X}, r')$;
- (ii) $\text{proj}_{\mathcal{X}}(\mathbf{x})$ is a singleton for all $\mathbf{x} \in U(\mathcal{X}, r')$;
- (iii) $\text{proj}_{\mathcal{X}}(\cdot)$ is p -Lipschitz continuous on $U(\mathcal{X}, r')$, where $p = r/(r-r')$;
- (iv) $N_{\text{px}}(\mathcal{X}, \cdot)$ is closed as a set-valued mapping;

$\text{proj}_{\mathcal{X}}(\mathbf{x}) : U(\mathcal{X}, r') \rightarrow \mathcal{X}$ well-defined and norm-to-norm continuous.

Informally speaking: local nonconvexities counterbalanced with smoothness of constraints



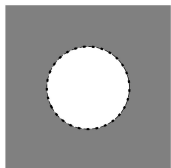
Federer, Curvature measures. TAMS, 1959.



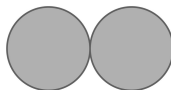
Nonempty closed convex set:
 ∞ -proximal smooth



Lack of proximal smoothness
("angle")



$\mathbb{R}^n \setminus B(0, r)$ is r -proximal
smooth



Non-proximal smooth union of
two convex sets

Sufficient conditions guaranteeing the prox-regularity for:

- ▷ A set defined by inequality/equality constraints
- ▷ Intersection of two prox-regular sets

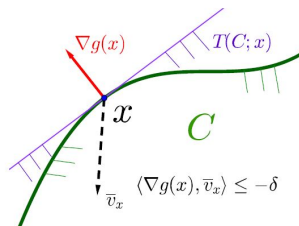
Differentiable and Lipschitz cont. inequality/equality on $U(\mathcal{X}, r)$

$$\text{(S.U.) } \exists \delta > 0, \forall x \in \text{bd}C, \exists \bar{v}_x \in \mathbb{B}, \forall k \in \{1, \dots, m\}, \forall \xi \in \partial_C g_k(x), \langle \xi, \bar{v}_x \rangle \leq -\delta$$

Single coupling constraint:
continuous differentiability only!



Adly, Thibault, Preservation of Prox-Regularity of Sets with Applications to Constrained Optimization. SIAM JO, 2016.



Assumptions 1 and 2 hold

Proposition: Existence

\mathcal{X} r -proximally smooth, for any $\gamma \in (0, \frac{r'}{1+F_{U(\mathcal{X}, r)}})$ with $r' \in (0, r)$:

- (i) $\mathbf{x}^* \in \mathcal{X}$ is a vCL-GNE;
- (ii) $\mathbf{x}^* = \text{proj}_{\mathcal{X}}(\mathbf{x}^* - \gamma F(\mathbf{x}^*))$, i.e., $\mathbf{x}^* \in \text{fix}(\text{proj}_{\mathcal{X}}(\text{Id} - \gamma F(\cdot)))$.

then the CL-GNEP has at least one vCL-GNE.

An additional assumption on the mapping F is required

Proposition: Local uniqueness

Mapping F is strictly monotone, then any vCL-GNE $\mathbf{x}^* \in \mathcal{X}$ is unique in its Clarke's tangent cone $\tilde{\mathcal{X}}(\mathbf{x}^*)$.

Equilibrium Computation

Assumption 3: strongly monotone pseudo-gradient mappings

\mathcal{X} is r -proximally smooth. Pseudo-gradient mapping F strongly monotone with $\mu > 0$ and Lipschitz continuous with $\ell > 0$.

$$(F(\mathbf{x}) - F(\mathbf{y}))^\top (\mathbf{x} - \mathbf{y}) \geq \mu \|\mathbf{x} - \mathbf{y}\|^2, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n$$

Algorithm 1: classical projected approach

$$\mathbf{x}^{k+1} = \text{proj}_{\mathcal{X}}(\mathbf{x}^k - \gamma F(\mathbf{x}^k))$$

Theorem: Existence and convergence

Let $\gamma \in (0, \min\{\frac{\sqrt{\ell^2 + p^2(\mu^2 - \ell^2)} + \mu p}{\ell^2 p}, \frac{r'}{1 + \bar{F}_U(\mathcal{X}, r)}\})$, $\mu \geq \ell$ and $r' \in (0, r)$.

- (i) The CL-GNEP has at least one vCL-GNE;
- (ii) The sequence generated by Algorithm 1 converges to a vCL-GNE.

To preserve the smoothness properties $\mathbf{x}^k - \gamma F(\mathbf{x}^k) \in U(\mathcal{X}, r')$

$$\begin{aligned} \left\| \mathbf{x}^{k+1} - \mathbf{x}^* \right\|^2 &= \left\| \text{proj}_{\mathcal{X}}(\mathbf{x} - \gamma F(\mathbf{x})) - \text{proj}_{\mathcal{X}}(\mathbf{x}^* - \gamma F(\mathbf{x}^*)) \right\|^2 \\ &\leq p^2 \left\| (\mathbf{x}^k - \gamma F(\mathbf{x}^k)) - (\mathbf{x}^* - \gamma F(\mathbf{x}^*)) \right\|^2 \\ &= p^2 \left\| (\mathbf{x}^k - \mathbf{x}^*) - \gamma (F(\mathbf{x}^k) - F(\mathbf{x}^*)) \right\|^2 \\ &= p^2 \left(\left\| \mathbf{x}^k - \mathbf{x}^* \right\|^2 + \gamma^2 \left\| F(\mathbf{x}^k) - F(\mathbf{x}^*) \right\|^2 - 2\gamma (F(\mathbf{x}^k) - F(\mathbf{x}^*))^\top (\mathbf{x}^k - \mathbf{x}^*) \right) \\ &\leq p^2 (1 - 2\gamma\mu + \gamma^2\ell^2) \left\| \mathbf{x}^k - \mathbf{x}^* \right\|^2 \end{aligned}$$

For proximally smooth sets the projection is Lipschitz continuous with constant $p = r/r'$.

Expansiveness of projection counterbalanced with the strong monotonicity of the pseudo-gradient mapping

Assumption 4: monotone pseudo-gradient mappings

\mathcal{X} is r -proximally smooth. F is monotone and Lipschitz continuous with constant $\ell > 0$.

$$(F(\mathbf{x}) - F(\mathbf{y}))^\top (\mathbf{x} - \mathbf{y}) \geq 0, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n$$

Algorithm 2: Modified version of the Korpelevich's approach.

$$\mathbf{x}^{k+1} = \text{proj}_{\tilde{\mathcal{X}}(\mathbf{y}^k)}(\mathbf{x}^k - \gamma F(\text{proj}_{\mathcal{X}}(\mathbf{x}^k - \gamma F(\mathbf{x}^k))))$$

Theorem: Existence and convergence

Let $\gamma \in (0, \min\{\frac{1}{\ell}, \frac{r}{4(1+\bar{F}_U(\mathcal{X}, r))}\})$ with $r' \in (0, r)$.

- (i) The CL-GNEP has at least one vCL-GNE;
- (ii) The sequence generated by Algorithm 2 converges to a vCL-GNE.

$$\begin{cases} \mathbf{y}^k = \text{proj}_{\mathcal{X}}(\mathbf{x}^k - \gamma F(\mathbf{x}^k)) \\ \mathbf{x}^{k+1} = \text{proj}_{\tilde{\mathcal{X}}(\mathbf{y}^k)}(\mathbf{x}^k - \gamma F(\mathbf{y}^k)) \end{cases}$$

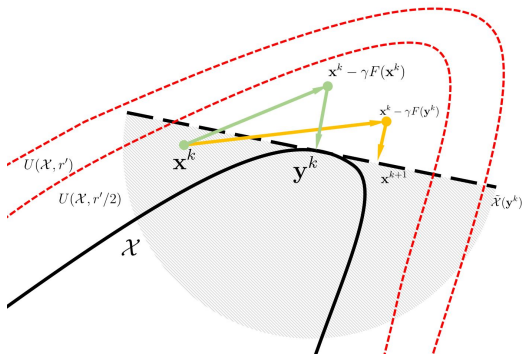
Each iteration

$\in U(\mathcal{X}, r')$.

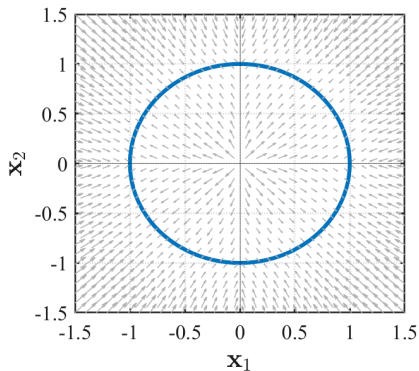
▷ $\mathbf{y}^k \in \mathcal{X}$

▷ $\mathbf{x}^k \in U(\mathcal{X}, r'/2)$.

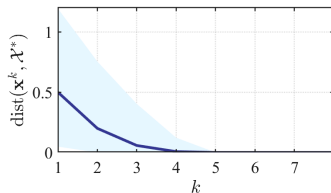
▷ $\mathbf{x}^k - \gamma F(\mathbf{x}^k) \in U(\mathcal{X}, r')$.



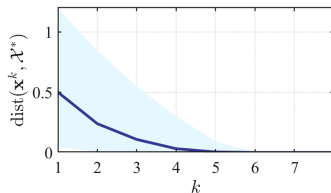
Example 1



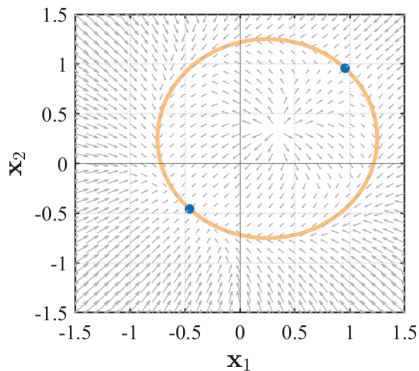
Algorithm 1



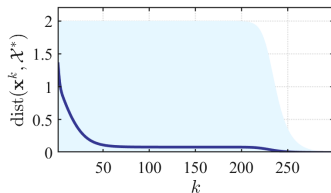
Algorithm 2



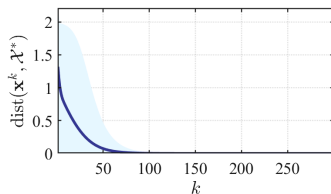
Example 2



Algorithm 1





Algorithm 2



Conclusion

- ▷ Definition of the problem ✓
- ▷ Optimality of the solution ✓
- ▷ Existence for proximally smooth sets ✓
- ▷ Uniqueness for proximally smooth sets ✓
- ▷ Convergence to a vCL-GNE on (strongly) monotone operators ✓
- ▷ Existence on more general settings ✗
- ▷ Distributed convergence ✗
- ▷ And so on...

 Scarabaggio, Carli, Grammatico & Dotoli. **Clarke's Local Equilibria in Nash Games with Nonconvex Coupling Constraints**. DOI: [10.36227/techrxiv.20079959](https://doi.org/10.36227/techrxiv.20079959)

 Scarabaggio, Carli & Dotoli. **Noncooperative Equilibrium Seeking in Distributed Energy Systems Under AC Power Flow Nonlinear Constraints**. *IEEE Transactions on Control of Network Systems*, 2022

Thanks for your attention!
Questions?