

Solving Multi-Leader-Follower Games

by Smoothing the Follower's Best Response

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Motivation

Toll Models

Toll Caps in Privatized Road Networks

by Harks, Schröder, and Vermeulen 2018

Electricity Markets

Nash equilibrium in a pay-as-bid electricity market

by Aussel, Bendotti, and Pištěk 2017

Analysis of m -stationary points to an epec modeling oligopolistic competition in an electricity spot market.

by Henrion, Outrata, and Surowiec 2012

- These models may include large amount of players
- Dynamic games get high dimensional
- Mean field limit transforms a high dimensional system of ODEs to a (nonlinear) PDE

Motivation: Example

Optimization and Mean Field Behavior for $N \rightarrow \infty$

$$\begin{aligned} & \text{for } i = 1, \dots, N : \\ & \min_{y_i, u_i} \int_0^T \phi(y_i, x) + \frac{\alpha}{2} u_i^2(t) dt \\ & \text{s.t. } \dot{y}_i(t) = u_i(t), y_i(0) = y_{i,0} \end{aligned}$$

$\xrightarrow{\text{MF}}$

$$\begin{aligned} & \min_{f, u} \int_0^T \int \left[\phi(y, x) + \frac{\alpha}{2} u^2 \right] f dy dt \\ & \text{s.t. } \partial_t f + \nabla_y (f u) = 0 \\ & f(0, y) = \frac{1}{N} \sum_{i=1}^N \delta(y - y_{i,0}) \end{aligned}$$

- Mean field limit can be applied on dynamic Multi-leader-follower games
- Relation between optimization and mean field limit
- Conditions for consistency

Existence (and Uniqueness) of equilibrium solutions for Multi-leader-follower games often even in finite dimensions unclear.

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Characterization of Nash Equilibria

Stationarity Concepts of MPCC

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General Games

N-Person Game: $\Gamma = \{\theta_\nu, X_\nu\}_{\nu=1}^N$

- Set $\{1, \dots, N\}$ of Players
- Strategy set $X_\nu \subseteq \mathbb{R}^{n_\nu}$ for each Player
- Pay-off function θ_ν for each player

$$\theta_\nu : X_1 \times X_2 \times \dots \times X_N \rightarrow \mathbb{R}$$



Optimization Problem



Nash Game



Stackelberg Game



Multi-Leader-Follower Game

Classification:

- Cooperative vs. **Noncooperative** Games
- **Continuous** vs. Discrete Games
- **Complete** vs. Incomplete Information

Nash Equilibrium (1950)

For all $\nu = 1, \dots, N$

$$\theta_\nu(x_\nu^*, x_{-\nu}^*) \leq \theta_\nu(x_\nu, x_{-\nu}^*)$$

for all $x_\nu \in X_\nu$.

$$x_{-\nu} = (x_1, \dots, x_{\nu-1}, x_{\nu+1}, \dots, x_N)$$

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Theoretical Results: Overview

For $\nu = 1, \dots, N$

MLFG

$$\begin{aligned} \min_{x_\nu} \quad & \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + a^\top y \\ \text{s.t.} \quad & x_\nu \in X_\nu \\ & \min_y \quad \frac{1}{2} y^\top Q_y y - b(x)^\top y \\ & \text{s.t.} \quad y \geq l(x) \end{aligned}$$

Convex NEP

$$\begin{aligned} \min_{x_\nu} \quad & \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + \sum_{i=1}^m a_i \max\{Q_y^{-1} b(x), l(x)\}_i \\ \text{s.t.} \quad & x_\nu \in X_\nu \text{ convex} \end{aligned}$$

Smooth NEP(ε)

$$\begin{aligned} \min_{x_\nu} \quad & \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + \frac{1}{2} \sum_{i=1}^m a_i [(L^\top + Q_y^{-1} B^\top) x + \tilde{\phi}_\varepsilon ((L^\top - Q_y^{-1} B^\top) x)]_i \\ \text{s.t.} \quad & x_\nu \in X_\nu \text{ convex} \end{aligned}$$

Theoretical Results: Existence

For $\nu = 1, \dots, N$

MLFG

$$\min_{x_\nu} \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + a^\top y$$

s.t. $x_\nu \in X_\nu$

$$\min_y \frac{1}{2} y^\top Q_y y - b(x)^\top y$$

s.t. $y \geq l(x)$



Convex NEP

$$\min_{x_\nu} \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + \sum_{i=1}^m a_i \max\{Q_y^{-1} b(x), l(x)\}_i$$

s.t. $x_\nu \in X_\nu$ convex

Assumptions (Follower)

- $Q_y \in \mathbb{R}^{m \times m}$ s.p.d. (and diagonal)
- $b_i, l_i : \mathbb{R}^n \rightarrow \mathbb{R}^m$ convex and smooth

$\Rightarrow \exists!$ (Nonsmooth) Best Response

$$y(x) = \max\{Q_y^{-1} b(x), l(x)\}$$

Assumptions (Leader)

- $Q_\nu \in \mathbb{R}^{n_\nu \times n_\nu}$ s.p.d.
- $a \in \mathbb{R}_+^m$
- $X_\nu \neq \emptyset$ convex and closed

Theorem (Existence)

The convex NEP (and therefore the MLFG) has at least one Nash equilibrium for compact X_ν .

Smooth Formulation

MLFG

$$\min_{x_\nu} \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + a^\top y$$

$$\text{s.t. } x_\nu \in X_\nu$$

$$\min_y \frac{1}{2} y^\top Q_y y - b(x)^\top y$$

$$\text{s.t. } y \geq l(x)$$

Follower's Best Response

$$l(x) = L^\top x, b(x) = B^\top x$$

■ Nonsmooth

$$y(x) = \max\{(Q_y^{-1} B^\top)x, L^\top x\}$$

■ Smooth

$$y_\varepsilon(x) = \frac{1}{2} \left[(L^\top + Q_y^{-1} B^\top) x + \underbrace{\tilde{\phi}_\varepsilon((L^\top - Q_y^{-1} B^\top) x)}_{\approx |(L^\top - Q_y^{-1} B^\top) x|} \right]$$

For $\nu = 1, \dots, N$

Smooth NEP(ε)

$$\min_{x_\nu} \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + \frac{1}{2} \sum_{i=1}^m a_i [(L^\top + Q_y^{-1} B^\top) x + \tilde{\phi}_\varepsilon((L^\top - Q_y^{-1} B^\top) x)]_i$$

$$\text{s.t. } x_\nu \in X_\nu \text{ convex}$$

Theorem: Existence and Uniqueness

The smoothed NEP with $\nu = 1, \dots, N$:

$$\min_{x_\nu} \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + \frac{1}{2} \sum_{i=1}^m a_i [(L^\top + Q_y^{-1} B^\top) x + \tilde{\phi}_\varepsilon ((L^\top - Q_y^{-1} B^\top) x)]_i$$

s.t. $x_\nu \in X_\nu$ convex and closed

has a **unique Nash equilibrium** for every smoothing parameter $\varepsilon > 0$.

Proof. We show that $\hat{\theta}' = (\nabla_{x_\nu} \theta_\nu(x_\nu, x_{-\nu}))_{\nu=1}^N$ is uniformly monotone since Q_ν s.p.d. and $\tilde{\phi}_\varepsilon \in C^2$ convex.

Therefore the VI($X, \hat{\theta}'$) has a unique solution.

Theoretical Results: Summary

For $\nu = 1, \dots, N$

MLFG

$$\begin{aligned} \min_{x_\nu} \quad & \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + a^\top y \\ \text{s.t.} \quad & x_\nu \in X_\nu \\ & \min_y \quad \frac{1}{2} y^\top Q_y y - b(x)^\top y \\ & \text{s.t.} \quad y \geq l(x) \end{aligned}$$

Convex NEP

$$\begin{aligned} \min_{x_\nu} \quad & \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + \sum_{i=1}^m a_i \max\{Q_y^{-1} b(x), l(x)\}_i \\ \text{s.t.} \quad & x_\nu \in X_\nu \text{ convex} \end{aligned}$$

Existence for
compact X_ν

Existence and
Uniqueness

Smooth NEP(ε)

$$\begin{aligned} \min_{x_\nu} \quad & \frac{1}{2} x_\nu^\top Q_\nu x_\nu + c_\nu^\top x_\nu + \frac{1}{2} \sum_{i=1}^m a_i [(L^\top + Q_y^{-1} B^\top) x + \tilde{\phi}_\varepsilon ((L^\top - Q_y^{-1} B^\top) x)]_i \\ \text{s.t.} \quad & x_\nu \in X_\nu \text{ convex} \end{aligned}$$

Extension to the Model

- Non-diagonal Q_y
 - auxiliary variable: $z = D^\top y$ where $Q_y = DD^\top$
 - $y(x) = D^{-\top} \max\{D^{-1}b(x), D^\top l(x)\}$
- Box Constraints $l(x) \leq y \leq u(x)$

$$y_i(x) = \text{median}(l_i(x), \tilde{y}_i(x), u_i(x)) = \begin{cases} \tilde{y}_i(x) & \text{else} \\ l_i(x) & \tilde{y}_i(x) \leq l_i(x) \\ u_i(x) & \tilde{y}_i(x) \geq u_i(x) \end{cases}$$

with $\tilde{y}_i(x) = (Q_y)_{ii}^{-1} b_i(x)$

- Multiple Follower
 - For $j = 1, \dots, N_F$ let:

$$\min_{y_j \in \mathbb{R}^{m_j}} \Theta_j(y_j, x) = \frac{1}{2} y_j^\top Q_y^j y_j - b_j(x)^\top y_j \quad \text{s.t.} \quad y_j \geq l_j(x)$$

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Characterization of NE: KKT(ε)

Joint KKT(ε) System if $X_\nu = \{x_\nu \in \mathbb{R}^{n_\nu} \mid g_\nu(x_\nu) \leq 0\}$ and assume that LICQ holds.

$$\begin{aligned}
 0 &= Qx + c_\nu^\top x_\nu + \frac{1}{2}(L^\top + Q_y^{-1}B^\top)^\top a \\
 &+ \frac{1}{2} \sum_{i=1}^m a_i (L^\top - Q_y^{-1}B^\top)^\top_{:,i} \tilde{\phi}'_\varepsilon'([(L^\top - Q_y^{-1}B^\top)x]_i) + \begin{bmatrix} \nabla_{x_{(1)}} g_{(1)}(x_{(1)}) \lambda_{(1)} \\ \vdots \\ \nabla_{x_{(N)}} g_{(N)}(x_{(N)}) \lambda_{(N)} \end{bmatrix} \\
 0 &= \min \left\{ \begin{bmatrix} \lambda_{(1)} \\ \vdots \\ \lambda_{(N)} \end{bmatrix}, \begin{bmatrix} -g_{(1)}(x_{(1)}) \\ \vdots \\ -g_{(N)}(x_{(N)}) \end{bmatrix} \right\}
 \end{aligned}$$

Lemma

Let $x^*(\varepsilon_k) \in X$ and $\lambda^*(\varepsilon_k) \in \Lambda$ for all $\varepsilon_k > 0$ and for X and Λ compact. Then there exists at least one accumulation point $x^*(0)$ and $\lambda^*(0)$ for a sequence $\varepsilon_k \rightarrow 0$.

Characterization of NE: KKT(0)

Fritz-John Conditions of Clarke

Let \bar{x} be a vector of local minimizer \bar{x}_ν of each leader problem; Slater condition holds for every X_ν , then there exists for $\nu = 1, \dots, N$ a multiplier $\lambda_\nu \in \mathbb{R}_+^{m_\nu}$:

$$0 \in Q_\nu \bar{x}_\nu + c_\nu^\top \bar{x}_\nu + \frac{1}{2} (L^\top + Q_\nu^{-1} B^\top)^\top a$$
$$+ \frac{1}{2} \sum_{i=1}^m a_i (L^\top - Q_\nu^{-1} B^\top)^\top_{\nu,i} \partial^C \tilde{\phi}_0 \left([(L^\top - Q_\nu^{-1} B^\top) \bar{x}]_i \right) + \nabla_{x_\nu} g_\nu(\bar{x}_\nu) \lambda_\nu$$
$$0 = \min \{ \lambda_\nu, -g_\nu(\bar{x}_\nu) \}$$

$$\partial^C \tilde{\phi}_0(z) = \begin{cases} 1, & z > 0 \\ -1, & z < 0 \\ [-1, 1], & z = 0 \end{cases}$$

$$\lim_{\varepsilon \rightarrow 0} \tilde{\phi}_\varepsilon'(z) = \begin{cases} 1, & z > 0 \\ -1, & z < 0 \\ 0, & z = 0 \end{cases}$$

The limit NE $x^*(0)$ satisfies a necessary optimality condition.

What about MPCC stationarity concepts?

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Mathematical Program with Complementarity Constraints (MPCC)

Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$, $g : \mathbb{R}^n \rightarrow \mathbb{R}^m$, and $G_1, G_2 : \mathbb{R}^n \rightarrow \mathbb{R}^l$ be smooth functions.

$$\begin{aligned} \min_z \quad & f(z) \\ \text{s.t.} \quad & g(z) \leq 0 \\ & 0 = \min \{G_1(z), G_2(z)\} \end{aligned} \tag{MPCC}$$

or alternatively: $0 \leq G_1(z) \perp G_2(z) \geq 0$

Stationarity MPCC

\bar{z} is a stationary point if there exist multipliers $(\lambda, \Gamma_1, \Gamma_2) \in \mathbb{R}^{m+l+l}$:

$$0 = \nabla_z f(\bar{z}) + \sum_{i=1}^m \lambda_i \nabla_z g_i(\bar{z}) - \sum_{i=1}^l \Gamma_{1,i} \nabla_z G_{1,i}(\bar{z}) - \sum_{i=1}^l \Gamma_{2,i} \nabla_z G_{2,i}(\bar{z})$$

$$0 \geq g(\bar{z}) \perp \lambda \geq 0$$

$$G_{1,i}(\bar{z}) \Gamma_{1,i} = 0, \quad i = 1, \dots, l$$

$$G_{2,i}(\bar{z}) \Gamma_{2,i} = 0, \quad i = 1, \dots, l$$

And for $i : G_{1,i}(\bar{z}) = G_{2,i}(\bar{z}) = 0$

S-Stationarity

$$\Gamma_{1,i} \geq 0 \quad \text{and} \\ \Gamma_{2,i} \geq 0$$

\Rightarrow M-Stationarity

$$\Gamma_{1,i}, \Gamma_{2,i} > 0 \quad \text{or} \\ \Gamma_{1,i} \cdot \Gamma_{2,i} = 0$$

\Rightarrow C-Stationarity

$$\Gamma_{1,i} \cdot \Gamma_{2,i} \geq 0$$

Leader Mathematical Program with Complementary Constraints

$$\begin{aligned} \min_{\mathbf{x}_\nu, y} \quad & \frac{1}{2} \mathbf{x}_\nu^\top Q_\nu \mathbf{x}_\nu + \mathbf{c}_\nu^\top \mathbf{x}_\nu + \mathbf{a}^\top y \\ \text{s.t.} \quad & g_\nu(\mathbf{x}_\nu) \leq 0 \\ & 0 = \min\{G_1(\mathbf{x}_\nu, \mathbf{x}_{-\nu}, y), G_2(\mathbf{x}_\nu, \mathbf{x}_{-\nu}, y)\} \end{aligned}$$

with $G_1(\mathbf{x}_\nu, \mathbf{x}_{-\nu}, y) = y - (Q_y^{-1} B^\top)^\top \mathbf{x}$ and $G_2(\mathbf{x}_\nu, \mathbf{x}_{-\nu}, y) = y - L^\top \mathbf{x}$.

Theorem

The limit $\mathbf{x}_\nu^*(0)$ is strongly stationary for the leader ν .

Proof. With the multipliers of the Fritz-John Conditions of Clarke we construct multipliers which fulfill the conditions of strong stationarity.

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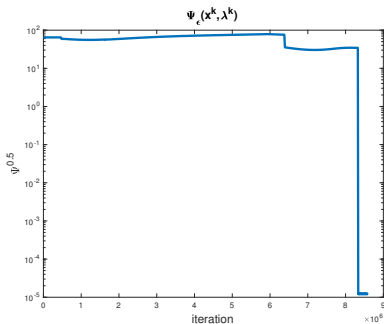
Computation

- Solution of $\text{KKT}(0)$?
- Solve sequence of $\Psi_\varepsilon(\mathbf{x}, \lambda) = \frac{1}{2} \|\text{KKT}(\varepsilon)\|_2^2$ for $\varepsilon_k \rightarrow 0$
- Ψ_ε is highly nonlinear and nonsmooth (but locally Lipschitz)
- Standard methods work slowly
- Importance to have good initial values

Computation: $\text{NEP}(\varepsilon)$

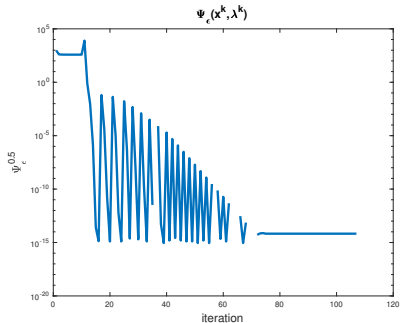
Subgradient

- first order derivatives
- characteristic convergence rate
- convergence guaranty for locally Lipschitz [Bagirov13]



Nonsmooth Newton

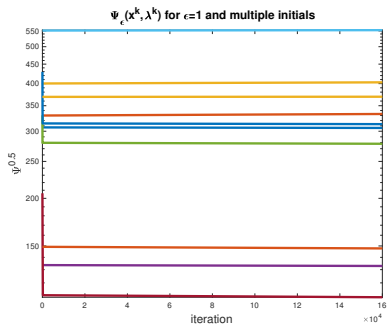
- second order derivatives
- very fast
- to be globalized by Subgradient method



Computation: $NEP(\varepsilon)$

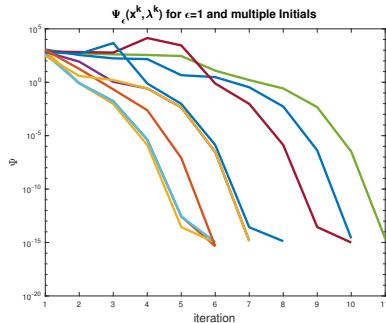
Subgradient

- first order derivatives
- characteristic convergence rate
- convergence guaranty for locally Lipschitz [Bagirov13]



Nonsmooth Newton

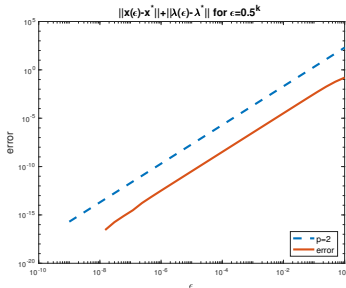
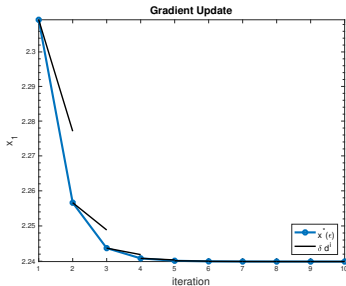
- second order derivatives
- very fast
- to be globalized by Subgradient method



Computation: Good Initials

Algorithm

1. Initialize $k = 0$, $z_0^0 = (x_0^0, \lambda_0^0)$, $\varepsilon_0 > 0$
2. Compute NE (x_k^*, λ_k^*) of $\text{NEP}(\varepsilon_k)$
3. Compute $\frac{\partial x}{\partial \varepsilon}$ by solving $\frac{d}{d\varepsilon} (\nabla_{x_\nu} \theta_\nu^\varepsilon(x_\nu(\varepsilon), x_{-\nu}(\varepsilon))) = 0$
4. choose $\varepsilon_{k+1} < \varepsilon_k$
5. $x_{k+1}^0 = x_k^* - (\varepsilon_k - \varepsilon_{k+1}) \frac{\partial x}{\partial \varepsilon}$
6. $\lambda_{k+1}^0 = \lambda_k^*$
7. $k++$; go to 2



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1. Quadratic Multi-Leader-Follower Game
2. NEP Formulation \Rightarrow Existence of NE for compact strategy sets
3. Smooth NEP(ε) Formulation \Rightarrow Existence&Uniqueness of NE(ε)
4. Limit NEP: $x^*(0)$ is strongly stationary
5. Efficient computation with globalized Nonsmooth Newton with $\frac{\partial x}{\partial \varepsilon}$ -based hot start

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Now: Multi-Level-Games with infinitely many players

For $k = 1, \dots, N_L$

$$\min_{\mathbf{x}_k} \int_0^T J_k^L(\mathbf{x}_k, \mathbf{x}_{-k}, m(y_1, \dots, y_{N_F})) dt$$

s.t. for $i = 1, \dots, N_F$:

$$\min_{y_i, u_i} \int_0^T \left[\phi(y_i, \mathbf{x}) + \frac{\alpha}{2} u_i^2 \right] dt$$

s.t. $\dot{y}_i = u_i, y_i(0) = y_{i,0}$

for $i = 1, \dots, N$:

$$\begin{aligned} \min_{y_i, u_i} \int_0^T \phi(y_i, x) + \frac{\alpha}{2} u_i^2(t) dt \\ \text{s.t. } \dot{y}_i(t) = u_i(t), y_i(0) = y_{i,0} \end{aligned}$$

$\xrightarrow{\text{MF}}$

$$\begin{aligned} \min_{f, u} \int_0^T \int [\phi(y, x) + \frac{\alpha}{2} u^2] f dy dt \\ \text{s.t. } \partial_t f + \nabla_y(fu) = 0 \\ f(0, y) = \frac{1}{N} \sum_{i=1}^N \delta(y - y_{i,0}) \end{aligned}$$

Outlook: Mean Field Limit

for $i = 1, \dots, N$:

$$\begin{aligned} \min_{y_i, u_i} \int_0^T \phi(y_i, x) + \frac{\alpha}{2} u_i^2(t) dt \\ \text{s.t. } \dot{y}_i(t) = u_i(t), y_i(0) = y_{i,0} \end{aligned}$$

- empirical measure

$$f(t, y) = \frac{1}{N} \sum_{i=1}^N \delta(y - y_i(t))$$

- compactly supported and smooth $\varphi(y)$
- $u(t, y_i(t)) = u_i(t)$

$$\begin{aligned} \frac{d}{dt} \int \varphi(y) f(t, y) dy &= \frac{1}{N} \sum_{i=1}^N \frac{d}{dt} \varphi(y) = \frac{1}{N} \sum_{i=1}^N \nabla_y \varphi(y_i(t)) \dot{y}_i(t) \\ &= \frac{1}{N} \sum_{i=1}^N \nabla_y \varphi(y_i(t)) u(t, y_i(t)) = \int \nabla_y \varphi(y) u(t, y) f(t, y) dy \\ &= - \int \nabla_y (u(t, y) f(t, y)) \varphi(y) dy \end{aligned}$$

$$\partial_t f + \nabla_y (f u) = 0$$

Thank you for your kind attention!

... and see you again at **ICCOPT** in Berlin in the session:
Nash Equilibrium Problems and Extensions

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