

## NON CONVEX QUADRATIC PROBLEMS

- We analyze the fundamental theory of Quadratic Problems (QP) **with single constraints**.
- These problems possess *strong duality* under Slater's condition.
- Quadratic Problem:

$$\begin{aligned} \min_x \quad & x^T A_0 x + 2b_0 x + c_0 \\ \text{s.t.} \quad & x^T A_1 x + 2b_1 x + c_1 \leq 0 \end{aligned} \quad (1)$$

where  $A_i \in \mathbf{S}^n$ ,  $b_i \in \mathbf{R}^n$ ,  $c_i \in \mathbf{R}$ .

- Lagrangian:  $L(x, \lambda) = x^T (A_0 + \lambda A_1) x + 2(b_0 + \lambda b_1)^T x + c_0 + \lambda c_1$
- Dual problem:

$$\begin{aligned} \max_{\lambda, \gamma} \quad & \gamma \\ \text{s.t.} \quad & \lambda \geq 0 \\ & \begin{bmatrix} A_0 + \lambda A_1 & b_0 + \lambda b_1 \\ (b_0 + \lambda b_1)^T & c_0 + \lambda c_1 - \gamma \end{bmatrix} \succeq 0 \end{aligned} \quad (2)$$

- Strong duality:** the optimal values of both problems coincide.

## POTENTIAL GAME

Consider a strategic non-cooperative game  $\mathcal{G} = \{\mathcal{Q}, \mathcal{X}, \{f_i\}_{i \in \mathcal{Q}}\}$  where

- $\mathcal{Q}$  is the set of  $Q$  players.
- $\mathcal{X} = \mathcal{X}_1 \times \dots \times \mathcal{X}_Q \subset \mathbf{R}^n$  is the set of pure strategies, i.e.  $x = (x_i)_{i \in \mathcal{Q}} \in \mathcal{Q}$ .
- function  $f_i : \mathcal{X}_i \rightarrow \mathbf{R}$  is the payoff for player  $i$ .
- A game  $\mathcal{G}$  is called an exact potential game if there exists a function  $V(x)$  such that

$$f_i(x_i, x_{-i}) - f_i(y_i, x_{-i}) = V(x_i, x_{-i}) - V(y_i, x_{-i}) \quad \forall x_i \in \mathcal{X}_i, x_{-i} \in \mathcal{X}_{-i}. \quad (4)$$

- EQUIVALENT QUADRATIC POTENTIAL PROBLEM:**

$$\begin{aligned} \min_x \quad & V(x) = x^T A_0 x + 2b_0^T x + \mathbf{1}_{n \times 1}^T c_0 \\ \text{s.t.} \quad & x_i^T A_1^i x_i + 2b_1^{iT} x_i + c_{1i} \leq 0 \quad \forall i \in \mathcal{Q}. \end{aligned} \quad (5)$$

- Potential problem has multiple constraints.**
- Solving the potential problem (5) provides an NE solution of the game.
- Notation:  $b_0 = (b_{0i})_{i=1}^Q$ ,  $b_1 = (b_{1i})_{i=1}^Q$ ,  $c_0 = (c_{0i})_{i=1}^Q$ ,  $A_1 = \text{diag}[A_1^1, \dots, A_1^1, \dots, A_1^Q]$ ,  $D(\lambda) = \text{diag}[\lambda] \otimes I_{n \times n}$ ,  $c_1 = (c_{1i})_{i=1}^Q$ , "diag" is the block diagonal matrix operator and " $\otimes$ " is the Kronecker product.

## ANALYSIS RESULTS OVER THE POTENTIAL PROBLEM

- Strong duality:** primal problem can be solved through the dual

$$q(\lambda) = \begin{cases} -(b_0 + D(\lambda)b_1)^T (A_0 + D(\lambda)A_1)^{\dagger} (b_0 + D(\lambda)b_1) \\ + \mathbf{1}_{n \times 1}^T c_0 + \lambda^T c_1 & \text{if } A_0 + D(\lambda)A_1 \succeq 0 \\ \text{and } (b_0 + D(\lambda)b_1) \in \mathcal{R}(A_0 + D(\lambda)A_1) \\ -\infty & \text{otherwise.} \end{cases}$$

- Coercivity:**  $\lim_{\|\lambda\| \rightarrow \infty} q(\lambda) \rightarrow -\infty$
- Existence of solution**  $\Leftrightarrow$  **existence of NE**  $\Leftrightarrow \{\lambda \in \mathbf{R}_+^Q \mid A_0 + D(\lambda)A_1 \succeq 0\}$  is nonempty.

## ALGORITHMS

- Centralized:** solve concave problem  $q(\lambda)$  and calculate

$$x^* \in -(A_0 + D(\lambda^*)A_1)^{\dagger} (b_0^T + b_1^T D(\lambda^*)) + \mathcal{N}(A_0 + D(\lambda^*)A_1)$$

where  $x^*$  is an NE of  $\mathcal{G}_p$ , and  $\mathcal{N}(Z)$  represents the nullspace of  $Z$ .

- Distributed:**

**Algorithm 1** Distributed Jacobi scheme ( $A_1^i \succ 0 \quad \forall i \in \mathcal{Q}$ )

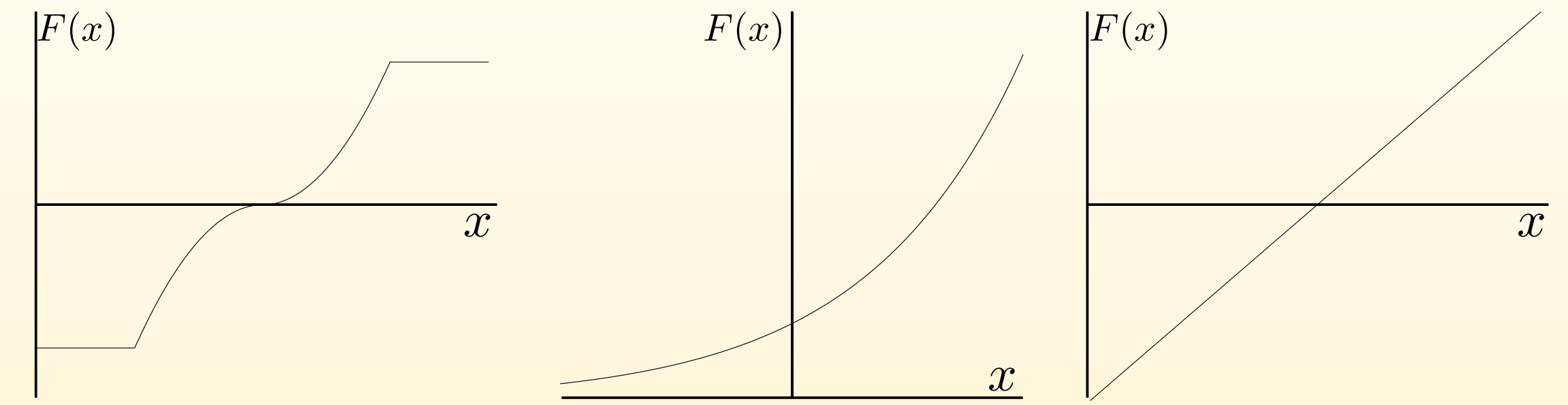
- Initialize  $(x_i^0)_i$ . Determine  $\lambda_i^{\min} \forall i$ . Set  $k \leftarrow 0$ .
- while**  $\|x^k - x^{k-1}\| \geq \varepsilon_{\text{outer}}$  **do**
- Set  $k \leftarrow k + 1$ .
- Calculate  $b_{gi} = b_{0i} + \sum_{j \neq i} A_0^{ij} x_j, \forall i$  //Mix strategies
- for**  $i \in \mathcal{Q}$  **do**
- Set  $\underline{\lambda}_i = \lambda_i^{\min}$ ,  $\bar{\lambda}_i = 2\lambda_i^{\min} + 1$ , and  $\bar{x}_i = \hat{x}_i(\bar{\lambda}_i, b_{gi})$ .
- while**  $h_i(\bar{x}_i) \geq 0$  **do** //Find bisection limits
- Update  $\Delta_i = \bar{\lambda}_i; \bar{\lambda}_i = 2\Delta_i$ . Solve  $\bar{x}_i = \hat{x}_i(\bar{\lambda}_i, b_{gi})$
- Set  $\Psi_{\text{cost}} \geq \varepsilon_{\text{inner}}$
- while**  $|\Psi_{\text{cost}}| \geq \varepsilon_{\text{inner}}$  **do** //Perform bisection steps
- Set  $\lambda_i^k = \frac{1}{2}(\bar{\lambda}_i + \underline{\lambda}_i)$ , determine  $x_i^k = \hat{x}_i(\lambda_i^k, b_{gi})$ .
- if**  $h_i(\bar{x}_i) \leq 0$ , **then**  $\bar{\lambda}_i = \lambda_i$
- else**,  $\underline{\lambda}_i = \lambda_i$ .
- if**  $\lambda_i^k > 0$ , **then**  $\Psi_{\text{cost}} = h_i(x_i^k)$  //Slackness violation
- else**,  $\Psi_{\text{cost}} = 0$  //case  $\lambda_i \approx 0$
- Solve  $(\lambda_i^k)_{i=1}^Q = \Pi_{\Gamma}((\lambda_i^k)_{i=1}^Q)$ , update  $x_i^k = \hat{x}_i(\lambda_i^k, b_{gi})$ .

## MONOTONICITY IN GAMES

- Given a convex subspace  $\mathcal{X} \subseteq \mathbf{R}^n$ , a mapping  $\mathbf{F} : \mathcal{X} \rightarrow \mathbf{R}^n$  is monotone if:

$$(\mathbf{F}(x) - \mathbf{F}(y))^T (x - y) \geq 0, \quad \forall x, y \in \mathcal{X} \quad (3)$$

- Examples (borrowed from [4]):



- Monotone games  $\Rightarrow$  existence of NE & algorithms that converge to NE.
- Coupling among players is limited.

- Monotonicity is a strong requirement.**

## PROBLEM FORMULATION

Given a set of players  $\mathcal{Q} = \{1, \dots, Q\}$ , we introduce the quadratic potential game  $\mathcal{G}_p$  where every player  $i \in \mathcal{Q}$  has to solve

$$\forall i \in \mathcal{Q} \quad \begin{cases} \min_{x_i \in \mathbf{R}^n} & f_i(x_i, x_{-i}) = x_i^T A_0^{ii} x_i + 2 \sum_{j \neq i} x_j^T A_0^{ij} x_i + 2b_{0i}^T x_i + c_{0i} \\ \text{s.t.} & h_i(x_i) = x_i^T A_1^i x_i + 2b_{1i}^T x_i + c_{1i} \leq 0 \end{cases} \quad (6)$$

- $A_0^{ii}, A_1^i \in \mathbf{S}^n, A_0^{ij} \in \mathbf{R}^{n \times n}$ ,  $\mathbf{S}^n$  is the set of symmetric matrices of size  $n$ ;
- $b_{0i}, b_{1i} \in \mathbf{R}^n$  are column vectors;
- $c_{0i}, c_{1i} \in \mathbf{R}$  are scalar numbers.
- The game is potential if, and only if, its Jacobian given by

$$A_0 = \begin{bmatrix} A_0^{11} & \dots & A_0^{1N} \\ \vdots & \ddots & \vdots \\ A_0^{N1} & \dots & A_0^{NN} \end{bmatrix},$$

is symmetric, i.e.,  $A_0^{ij} = (A_0^{ji})^T$ .

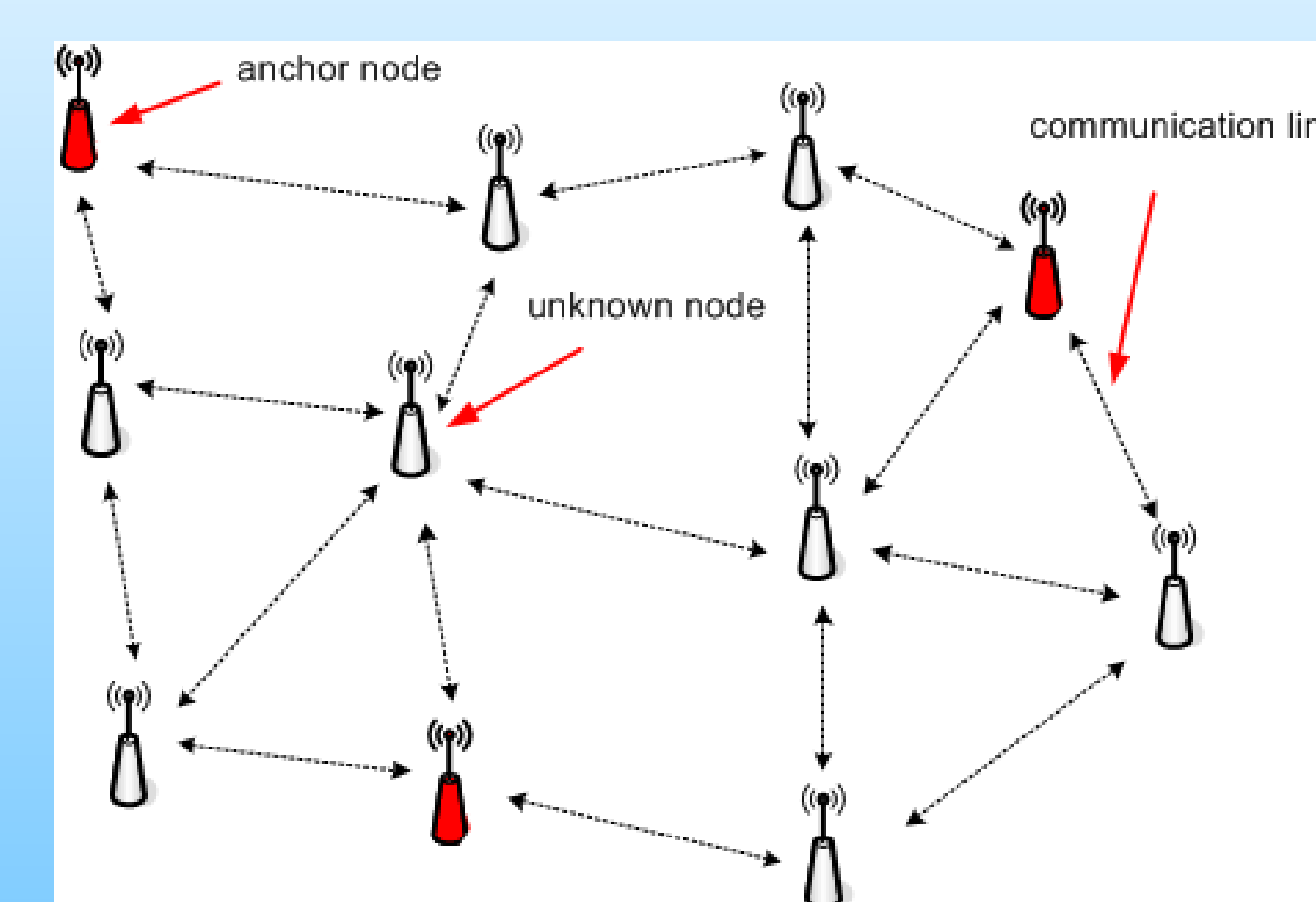
**$A_0^{ii}, A_1^i$  do not need to be positive semidefinite. The problems do not need to be convex.**

**The quadratic game does not need to be monotone.**

## APPLICATIONS

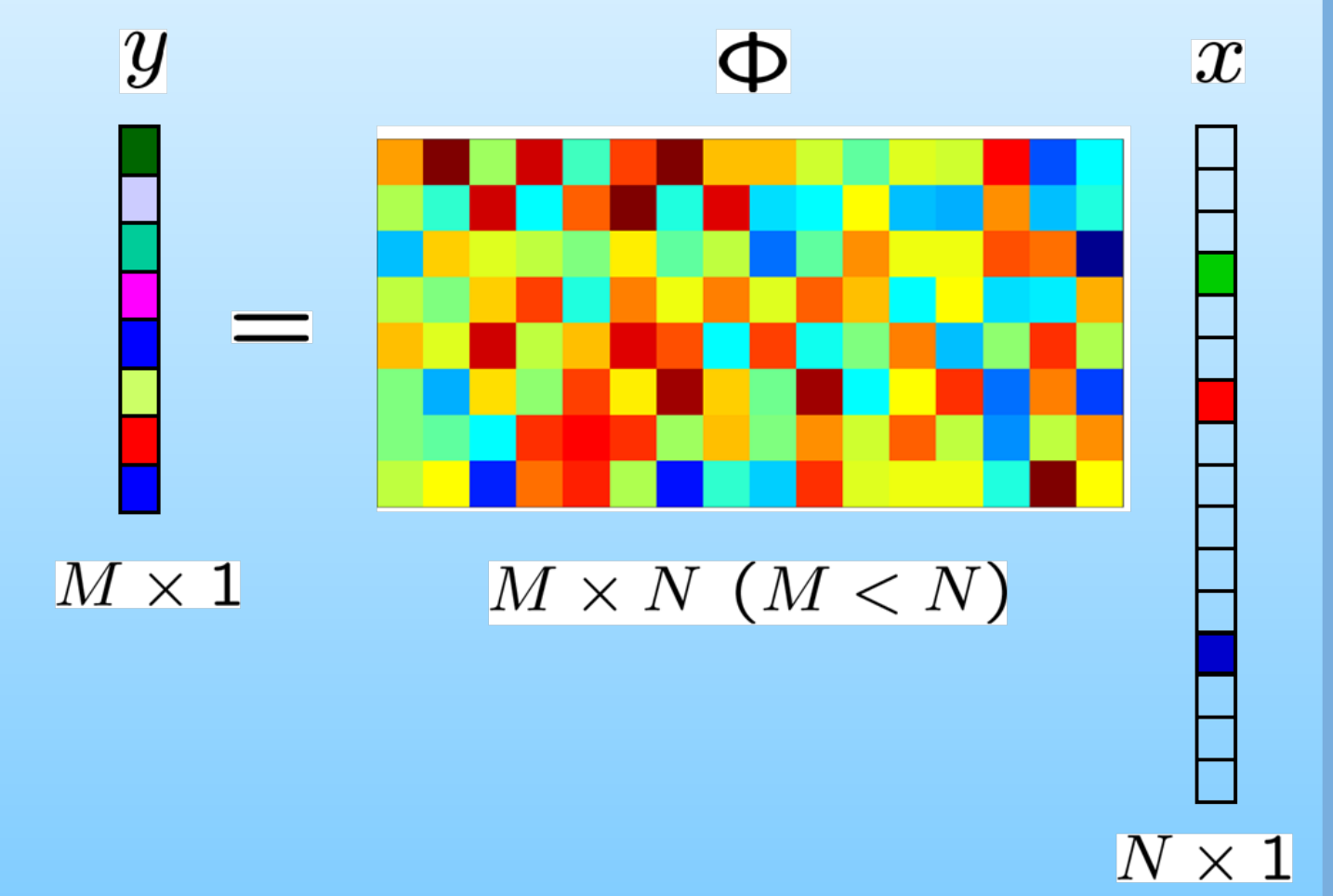
- Optimal localization (Non-Convex)

$$\min_{x \in \mathbf{R}^n} \sum_{i \in \mathcal{Q}} (d_i^2 - \|x - y_i\|^2)^2$$



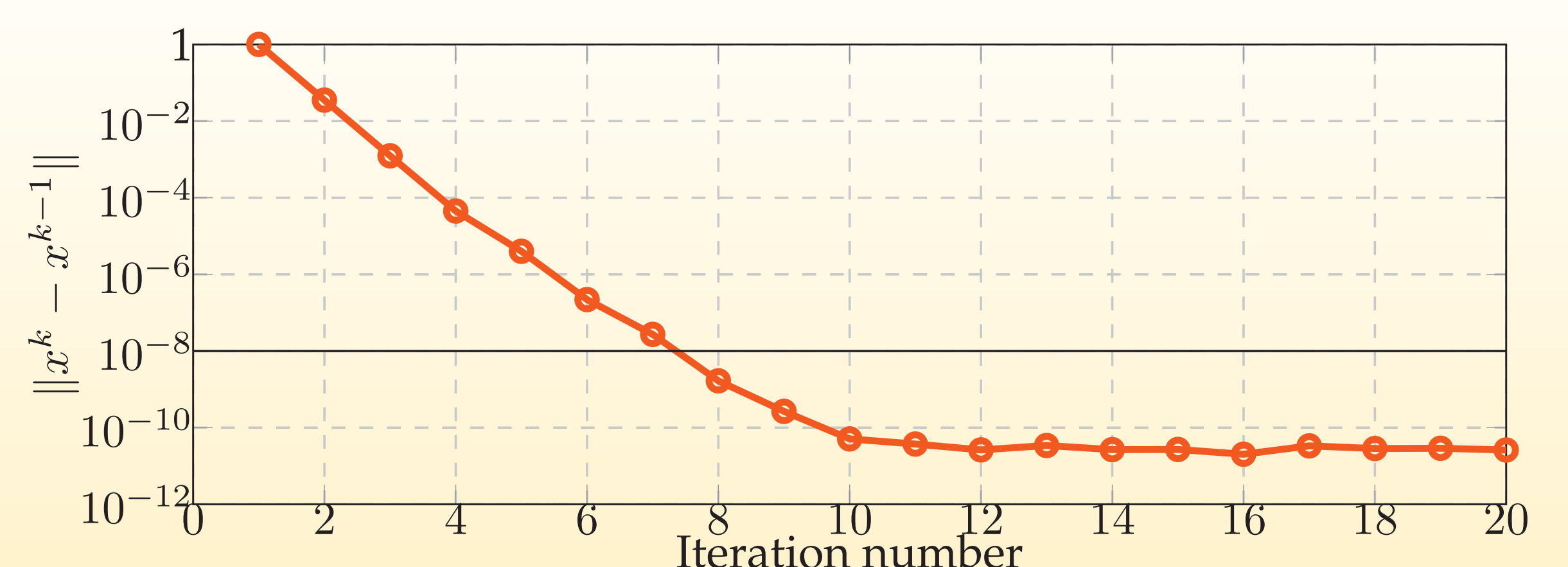
- Robust Least Squares (MinMax)

$$\min_{x \in \mathbf{R}^n} \max_{\{(\Delta_i, \delta_i) \mid \|\Delta_i\| \leq \Gamma_i\}} \|(A + \Delta)x - \delta - b\|$$



## SIMULATIONS

- 200 simulated games,  $Q = 10$  payers, and  $n = 4$ .



## BASIC REFERENCES

- S. Boyd and L. Vandenberghe, Convex optimization, Cambridge University Press, 2004.
- V. Jeyakumar, A.M. Rubinov, and Z.Y. Wu, "Non-convex quadratic minimization problems with quadratic constraints: global optimality conditions," Mathematical Programming, vol. 110, no. 3, pp. 521-541, Aug. 2006.
- F. Facchinei, V. Piccialli, and M. Sciandrone, "Decomposition algorithms for generalized potential games," Computational Optimization and Applications, vol. 50, no. 2, pp. 237-262, Oct. 2011.
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