## Stability in Matrix Games


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## Main idea

Classical settings. Matrix games and Linear Programming (LP). Classical question. Stability:

How do our objects of interest change upon perturbations?
Observables. Solutions and value of the problems.

## How do solutions and value change upon perturbations?

## Matrix Games

$$
i\binom{j}{m_{i, j}}
$$

$$
\operatorname{val} M:=\max _{p \in \Delta[m]} \min _{q \in \Delta[n]} p^{t} M q
$$

$$
M(\varepsilon)=M_{0}+M_{1} \varepsilon
$$

## Derivative of the value function [Mills56]

Define

$$
D \operatorname{val} M\left(0^{+}\right):=\lim _{\varepsilon \rightarrow 0^{+}} \frac{\operatorname{val} M(\varepsilon)-\operatorname{val} M(0)}{\varepsilon}
$$

## Results.

(1) Characterization of $\operatorname{Dval} M\left(0^{+}\right)$.
(2) (Poly-time) algorithm for computing it.

## Theorem ([Milis56])

Given $M(\varepsilon)=M_{0}+M_{1} \varepsilon$,

$$
D \operatorname{val} M\left(0^{+}\right)=\operatorname{val}_{P\left(M_{0}\right) \times Q\left(M_{0}\right)} M_{1} .
$$

## Our framework

Polynomial matrix games. Matrix games where payoff entries are given by polynomials.

$$
M(\varepsilon)=M_{0}+M_{1} \varepsilon+\ldots+M_{K} \varepsilon^{K} .
$$

Definition (Value-positivity problem)
$\exists \varepsilon_{0}>0$ such that $\forall \varepsilon \in\left[0, \varepsilon_{0}\right] \quad \operatorname{val} M(\varepsilon) \geq \operatorname{val} M(0)$.

## Definition (Uniform value-positivity problem)

$\exists p_{0} \in \Delta[m] \quad \exists \varepsilon_{0}>0 \quad \forall \varepsilon \in\left[0, \varepsilon_{0}\right] \quad \operatorname{val}\left(M(\varepsilon) ; p_{0}\right) \geq \operatorname{val} M(0)$.

Definition (Functional form problem)
Return the maps $\operatorname{val} M(\cdot)$ and $p^{*}(\cdot)$, for $\varepsilon \in\left[0, \varepsilon_{0}\right]$.

## Polynomial matrix game

Consider $\varepsilon>0$.

$$
M(\varepsilon)=\left(\begin{array}{cc}
1 & -1 \\
-1 & 1
\end{array}\right)+\left(\begin{array}{cc}
1 & -3 \\
0 & 2
\end{array}\right) \varepsilon
$$

The optimal strategy is given by, for $\varepsilon<1 / 2$,

$$
p_{\varepsilon}^{*}=\left(\frac{1+\varepsilon}{2+3 \varepsilon}, \frac{1+2 \varepsilon}{2+3 \varepsilon}\right)^{t}
$$

Therefore,

$$
\operatorname{val} M(\varepsilon)=\frac{\varepsilon^{2}}{2+3 \varepsilon}
$$

## Polynomial matrix game, negative direction

Consider $\varepsilon>0$.

$$
M(\varepsilon)=\left(\begin{array}{cc}
1 & -1 \\
-1 & 1
\end{array}\right)+\left(\begin{array}{cc}
-1 & 3 \\
0 & -2
\end{array}\right) \varepsilon
$$

The optimal strategy is given by, for $\varepsilon<2 / 3$,

$$
p_{\varepsilon}^{*}=\left(\frac{1-\varepsilon}{2-3 \varepsilon}, \frac{1-2 \varepsilon}{2-3 \varepsilon}\right)^{t} .
$$

Therefore,

$$
\operatorname{val} M(\varepsilon)=\frac{\varepsilon^{2}}{2-3 \varepsilon}
$$

## Types of questions

## Question (Complexity)

Determining the complexity of solving these three problems.

## Question (Equivalent problem)

Finding the equivalent problems in the setting of Linear Programming.

## Algorithms

## Lemma (Poly-time algorithms)

There are certifying polynomial-time algorithms for all three problems, for rational data.

## Remark

This extends the work of [Mills56].

## Equivalence with LPs

## Theorem (Adler03)

Matrix games and LPs a poly-time equivalent.

- [Dantzig51] gives an incomplete proof.
- The reduction depends on the computational model: rational, algebraic or real data.


## LPs with errors

An error-free LP is the following optimization problem.

$$
\left(P_{0}\right)\left\{\begin{array}{lrl}
\min _{x} & c_{0}^{t} x & \\
\text { s.t. } & A_{0} x & \leq b_{0} \\
& x & \geq 0
\end{array}\right.
$$

An LP with errors considers polynomials $A(\varepsilon), b(\varepsilon), c(\varepsilon)$.

## LP with errors

$$
\begin{aligned}
& \leftrightarrow \\
& \left(P_{\varepsilon}\right)\left\{\begin{array}{lrl}
\min _{x} & x & \\
\text { s.t. } & x & \leq-\varepsilon \\
& -x & \leq-\varepsilon .
\end{array}\right.
\end{aligned}
$$

It is not weakly robust, therefore not strongly robust and there is no functional form.

## LP with errors

It is weakly robust and strongly robust. Therefore, for $\varepsilon<1$, the functional form is:

$$
\begin{aligned}
\operatorname{val}\left(P_{\varepsilon}\right) & \equiv 0 \\
(x, y)^{*}(\varepsilon) & \equiv(0,0)
\end{aligned}
$$



## Mills in LPs

## Theorem ([Mills56])

Under some regularity conditions, the derivative of the value of an $L P$ has a similar characterization as the matrix game.

## Asymptotic optimality

## Definition (Asymptotically optimal)

A function $x^{*}$ that is optimal for all $\varepsilon \in\left(0, \varepsilon_{0}\right]$.

## Lemma (AFJ13)

There is an EXPTIME algorithm to compute an asymptotically optimal solution of an LP with errors.

## Remark

It consists in a variation of the Simplex method.

## Reduction

## Definition (Weakly robust)

$\exists \varepsilon_{0}>0$ such that, $\forall \varepsilon \in\left[0, \varepsilon_{0}\right] \quad\left(P_{\varepsilon}\right)$ is feasible and bounded.
Definition (Strongly robust)
$\exists x^{*} \quad \exists \varepsilon_{0}>0 \quad \forall \varepsilon \in\left[0, \varepsilon_{0}\right], \quad x^{*}$ is also a solution of $\left(P_{\varepsilon}\right)$.
Definition (Functional form)
The maps $\operatorname{val}(P$.$) and x^{*}(\cdot)$, for $\varepsilon \in\left[0, \varepsilon_{0}\right]$.

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Lemma (Reduction from LP with error to polynomial matrix games)
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There is a polynomial-time reduction from robustness problems to the respective value-positivity problem, which preserves the degree of the error perturbation, for algebraic data.

## Types of questions (again)

## Question (Complexity)

Determining the complexity of solving these three problems.

## Question (Equivalent problem)

Finding the equivalent problems in the setting of Linear Programming.

## Poly-time for value-positivity

## Lemma (Shapley and Snow kernel [SS52])

Let $M$ be a matrix game. There exists a square submatrix $\dot{M}$ such that $\mathbb{1}^{t} c o(\dot{M}) \mathbb{1} \neq 0$ and

$$
\operatorname{val} M=\frac{\operatorname{det} \dot{M}}{\mathbb{1}^{t} \operatorname{co}(\dot{M}) \mathbb{1}}
$$

where $\operatorname{co}(\dot{M})$ is the co-matrix of $\dot{M}$ and $\mathbb{1}$ is the juector of ones.


## LP reduction

Follows from the work of [Adler13]:

- Given an LP with errors, apply the reduction.
- Solving value-positivity problems to the resulting matrix games answers robustness questions.


## Comments



## Zero-sum discounted stochastic games

## Definition (Stochastic game)

A collection of matrix games $M_{1}, M_{2}, \ldots, M_{K}$ together with a transition function $q$ and a discount factor $\lambda \in(0,1]$. The payoff is given by

$$
\lambda \sum_{i \geq 0}(1-\lambda)^{i} R_{i}
$$

## Towards zero-sum discounted stochastic games

Recall our results about polynomial matrix games.

- [Mills56] presents the idea of iteratively solving a game to find the derivative.
- This is true in more general context than matrix games [MSZ15].
- Perturbation of matrix games has been extended to discounted zero-sum stochastic games [TV80, AFFG01].


## Question

Can we extend our analysis to zero-sum discounted stochastic games?

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[^0]:    - Parametrized flows
    - Derivable matrix games?

