

TWO-PLAYER ZERO-SUM GAMES

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HOW MORRA IS PLAYED

- Each player raises between 1 and 3 fingers and simultaneously makes a guess about how many fingers the opponent will raise
- There is no payoff unless *exactly one player predicts correctly*
- The correct guesser wins an amount from the other player, which is equal to the total number of fingers raised by both players

THE PAYOFF MATRIX FOR MORRA

	1-1	1-2	1-3	2-1	2-2	2-3	3-1	3-2	3-3
1-1	0	2	2	-3	0	0	-4	0	0
1-2	-2	0	0	0	3	3	-4	0	0
1-3	-2	0	0	-3	0	0	0	4	4
2-1	3	0	3	0	-4	0	0	-5	0
2-2	0	-3	0	4	0	4	0	-5	0
2-3	0	-3	0	0	-4	0	5	0	5
3-1	4	4	0	0	0	-5	0	0	-6
3-2	0	0	-4	5	5	0	0	0	-6
3-3	0	0	-4	0	0	-5	6	6	0

TWO-PLAYER ZERO-SUM GAMES

Two-player zero-sum game (equivalently, matrix game) is given by

1. Player set $N = \{1, 2\}$
2. Finite strategy sets S_1 and S_2
3. Utility functions satisfying $u_1 + u_2 = 0$

Remarks

- Notation $u := u_1 = -u_2$
- We can view $u(s_1, s_2)$ as the payoff of player 1/loss of player 2
- Terminology: player 1 is *maximizing* while player 2 is *minimizing*

MINIMAX THEOREM

SOLVING MATRIX GAMES

- By the Nash theorem, any matrix game has an equilibrium $(p_1^*, p_2^*) \in \Delta$ in mixed strategies,

$$U(p_1, p_2^*) \leq U(p_1^*, p_2^*) \leq U(p_1^*, p_2) \quad \forall (p_1, p_2) \in \Delta$$

- We derive this result from fundamental principles, which will lead naturally to a linear programming (LP) problem

PURE EQUILIBRIA IN MATRIX GAMES

- A pure Nash equilibrium in a matrix game is a pair $(s_1^*, s_2^*) \in \mathbf{S}$ s.t.

$$u(s_1, s_2^*) \leq u(s_1^*, s_2^*) \leq u(s_1^*, s_2) \quad \forall (s_1, s_2) \in \mathbf{S}$$

- The strategy profile (s_1^*, s_2^*) is also called a **saddle point**

	<i>d</i>	<i>e</i>	<i>f</i>
<i>a</i>	1	2	3
<i>b</i>	4	5	6
<i>c</i>	7	8	9

$$\max_{s_1 \in S_1} \min_{s_2 \in S_2} u(s_1, s_2) = 7 = \min_{s_2 \in S_2} \max_{s_1 \in S_1} u(s_1, s_2)$$

MAXIMIN/MINIMAX VALUE OF A MATRIX GAME

Lower bound on the payoff

1. Given p_1 , player 2 computes

$$\min_{p_2 \in \Delta_2} U(p_1, p_2)$$

2. Player 1 then computes

$$\begin{aligned} \underline{v} &:= \max_{p_1 \in \Delta_1} \min_{p_2 \in \Delta_2} U(p_1, p_2) \\ &= \max_{p_1 \in \Delta_1} \min_{s_2 \in S_2} U(p_1, s_2) \end{aligned}$$

Upper bound on the loss

1. Given p_2 , player 1 computes

$$\max_{p_1 \in \Delta_1} U(p_1, p_2)$$

2. Player 2 then computes

$$\begin{aligned} \bar{v} &:= \min_{p_2 \in \Delta_2} \max_{p_1 \in \Delta_1} U(p_1, p_2) \\ &= \min_{p_2 \in \Delta_2} \max_{s_1 \in S_1} U(s_1, p_2) \end{aligned}$$

LP FORMULATION

Player 1 solves

$$\max_{p_1 \in \Delta_1} \min_{s_2 \in S_2} U(p_1, s_2)$$

Maximize v_1

subject to

$$U(p_1, s_2) \geq v_1 \quad \forall s_2 \in S_2$$

$$p_1 \in \Delta_1$$

$$v_1 \in \mathbb{R}$$

Player 2 solves

$$\min_{p_2 \in \Delta_2} \max_{s_1 \in S_1} U(s_1, p_2)$$

Minimize v_2

subject to

$$U(s_1, p_2) \leq v_2 \quad \forall s_1 \in S_1$$

$$p_2 \in \Delta_2$$

$$v_2 \in \mathbb{R}$$

Minimax theorem (von Neumann, 1928)

The two LPs are dual and their optimal value is $v := \underline{v} = \bar{v}$, which is called the **value** of the game.

NASH EQUILIBRIA IN MATRIX GAMES

- **Maximin strategy** is the optimal solution p_1^* for player 1
- **Minimax strategy** is the optimal solution p_2^* for player 2

Proposition

Let (p_1^, p_2^*) be a mixed strategy profile in a matrix game. The following are equivalent.*

- 1. p_1^* is a maximin strategy and p_2^* is a minimax strategy.*
- 2. (p_1^*, p_2^*) is a Nash equilibrium.*

If any of the above conditions hold, then $v = U(p_1^, p_2^*)$.*

THE SOLUTION OF MORRA

	1-1	1-2	1-3	2-1	2-2	2-3	3-1	3-2	3-3
1-1	0	2	2	-3	0	0	-4	0	0
1-2	-2	0	0	0	3	3	-4	0	0
1-3	-2	0	0	-3	0	0	0	4	4
2-1	3	0	3	0	-4	0	0	-5	0
2-2	0	-3	0	4	0	4	0	-5	0
2-3	0	-3	0	0	-4	0	5	0	5
3-1	4	4	0	0	0	-5	0	0	-6
3-2	0	0	-4	5	5	0	0	0	-6
3-3	0	0	-4	0	0	-5	6	6	0

$$v = 0$$

The support of any maximin strategy is $\{1-3, 2-2, 3-1\}$, e.g.

$$p_{13}^* = \frac{5}{12}, p_{22}^* = \frac{4}{12}, p_{31}^* = \frac{3}{12}$$

Maximize v_1

subject to

$$-2p_{12} - 2p_{13} + 3p_{21} + 4p_{31} \geq v_1$$

$\vdots$

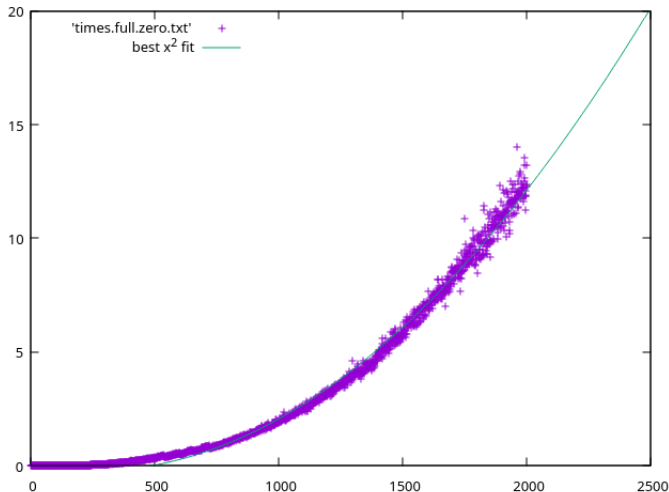
$$p_{ij} \geq 0 \quad i, j = 1, 2, 3$$

$$\sum_{i=1}^3 \sum_{j=1}^3 p_{ij} = 1$$

$$v_1 \in \mathbb{R}$$

COMPUTATIONAL EXPERIMENTS

- Julia + JuMP + Gurobi, randomly generated matrix games
- Number of strategies vs Solve time in Gurobi



COMPARISON

	<i>Zero-sum</i>	<i>General-sum</i>
Nash equilibrium	exists	exists
maxmin/minmax strategies	equivalent to NE	different
unique value	yes	no
equilibrium selection problem	no	yes
computable in $\mathbb{Q}$	yes	no
optimization problem	LP	non-convex POP

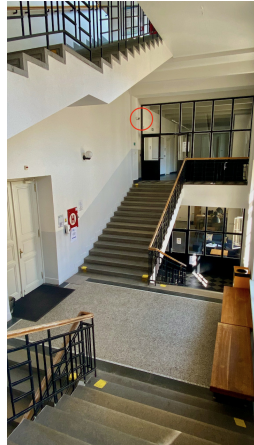
DOUBLE ORACLE ALGORITHM

MOTIVATION

- Certain matrix games are too large to solve directly using the baseline linear programming approach
- We will discuss [strategy generation](#) method which gradually expands the sets of currently used strategies

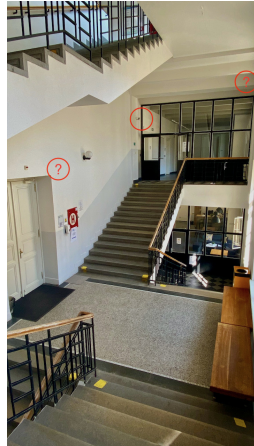
WHAT PATH SHOULD THE ROBOT FOLLOW TO AVOID CCTV?

The position of cameras is **known**.



WHAT PATH SHOULD THE ROBOT FOLLOW TO AVOID CCTV?

The **adversary** deploys cameras.



WHAT PATH SHOULD THE ROBOT FOLLOW TO AVOID CCTV?

Motion planner

- Path π for the robot
- Finite set of paths Π
- Mixed strategy $p \in \Delta_{\Pi}$
- Loss $\ell(\pi, \mathbf{c})$
- Expected loss

$$\sum_{\pi \in \Pi} \sum_{\mathbf{c} \in C} p(\pi) \cdot q(\mathbf{c}) \cdot \ell(\pi, \mathbf{c})$$

Adversary

- Cost vector $\mathbf{c}$
- Finite set of cost vectors C
- Mixed strategy $q \in \Delta_C$

PLANNING PATHS: EXPERIMENTS

McMahan, Gordon, Blum (ICML 2003)

- The gridworld of size up to 269×226
- The robot can move in any of 16 compass directions
- Each cell has cost 1 and a cost proportional to the distance of camera

Computational limits

- Sets Π and C should be reasonably small
- Already $\binom{100}{2} = 4\,950$ positions for 2 cameras in the gridworld 10×10

DOUBLE ORACLE ALGORITHM

Input: Any matrix game with strategy sets S_1 and S_2

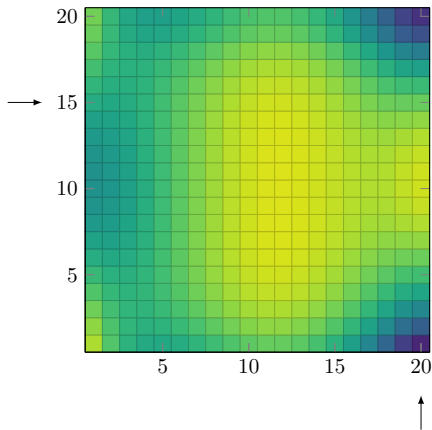
Initialize: Pick small strategy sets $T_1 \subseteq S_1$ and $T_2 \subseteq S_2$

1. Solve the subgame with T_1 and $T_2 \longrightarrow (q_1^*, q_2^*)$
2. Compute the pure best responses

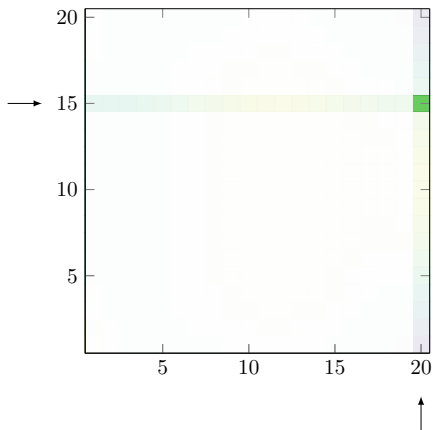
$$s_1 \in \operatorname{argmax}_{s'_1 \in S_1} U(s'_1, q_2^*) \quad \text{and} \quad s_2 \in \operatorname{argmin}_{s'_2 \in S_2} U(q_1^*, s'_2)$$

3. If $s_1 \in T_1$ and $s_2 \in T_2$, then stop
4. Otherwise add s_1 to T_1 or s_2 to T_2 , and go to 1.

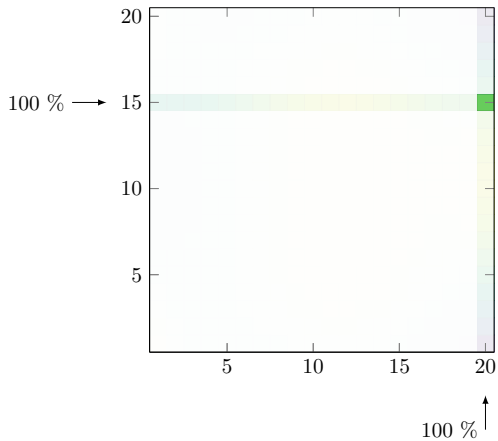
DOUBLE ORACLE ALGORITHM



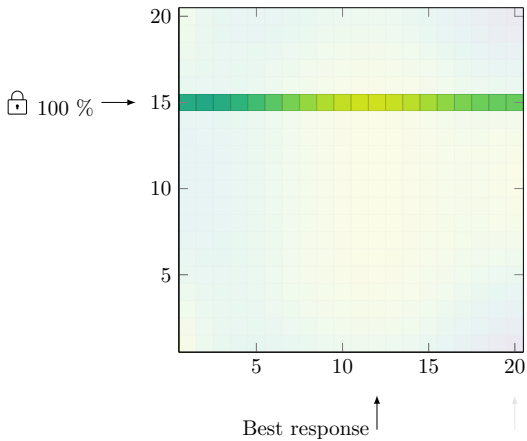
DOUBLE ORACLE ALGORITHM



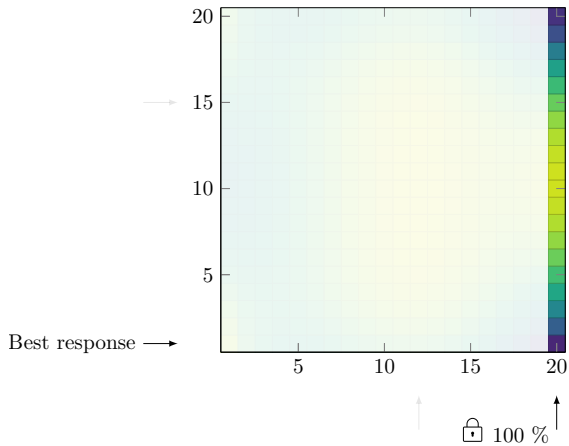
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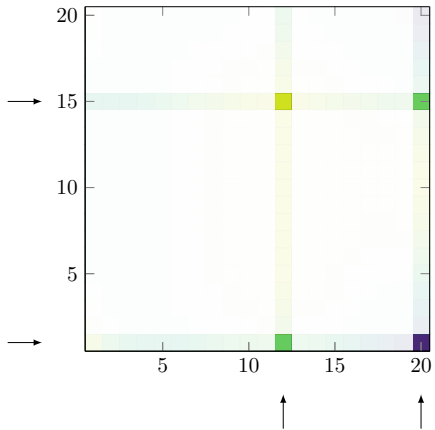
DOUBLE ORACLE ALGORITHM



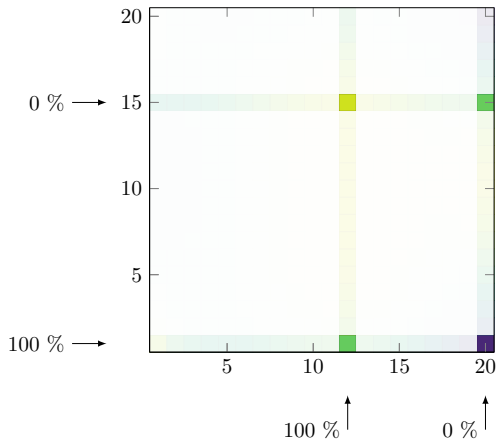
DOUBLE ORACLE ALGORITHM



DOUBLE ORACLE ALGORITHM



DOUBLE ORACLE ALGORITHM



TERMINATION AND CORRECTNESS

Proposition

The DO algorithm terminates and returns a Nash equilibrium of the initial matrix game.

In each iteration:

1. $v \leq v_u := U(s_1, q_2^*)$
2. $v \geq v_\ell := U(q_1^*, s_2)$
3. If $s_1 \in T_1$ and $s_2 \in T_2$, then $v_\ell = v = v_u$ and (q_1^*, q_2^*) is a NE

ALTERNATIVE TERMINATING CONDITION

- Choose some $\varepsilon > 0$ and stop when

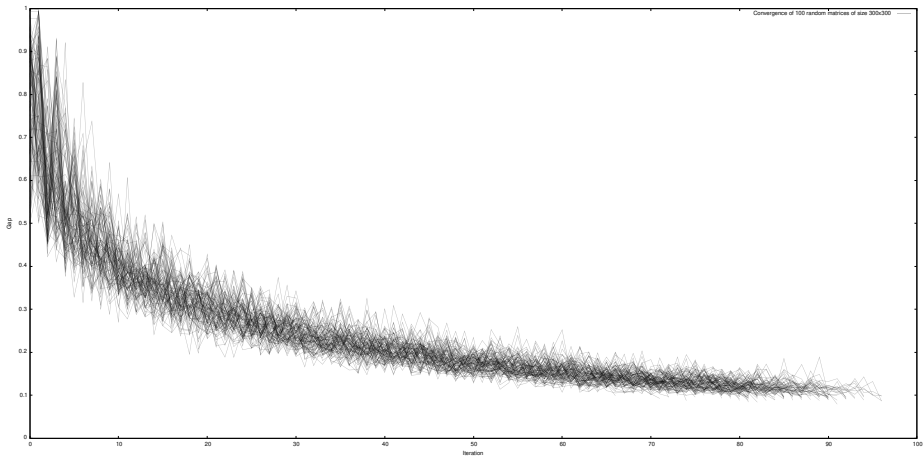
$$V_U - V_\ell \leq \varepsilon$$

- The output (q_1^*, q_2^*) is an ε -Nash equilibrium,

$$U(p_1, q_2^*) - \varepsilon \leq U(q_1^*, q_2^*) \leq U(q_1^*, p_2) + \varepsilon \quad \forall (p_1, p_2) \in \Delta$$

CONVERGENCE TO ε -EQUILIBRIUM

#iterations vs convergence criterion for 300×300 games



CONVERGENCE TO ε -EQUILIBRIUM

