

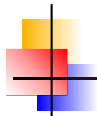


An Approximate Subgame-Perfect Equilibrium Computation Technique for Repeated Games

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Plan

Motivation

Game Theory Background

Problem and Approach

Conclusion and Future Work



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Motivation

- ▶ Discover an algorithmic way for:
 - ▶ Finding equilibrium solutions for dynamic games
 - ▶ Computing equilibrium strategies for dynamic game players

▶ Prisoner's Dilemma

		Player 2	
		<i>C</i>	<i>D</i>
Player 1	<i>C</i>	2, 2	-1, 4
	<i>D</i>	4, -1	0, 0

- ▶ When the discount factor is close enough to 1, the long-term average payoff profile (2, 2) is an equilibrium point and there is a strategy, which each player can adopt for generating that point: Tit-For-Tat
- ▶ For an arbitrary discount factor, we don't usually know:
 - ▶ What is the set of equilibrium points?
 - ▶ What are the strategies of players that generate those equilibrium points?



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- ▶ A stage-game is a tuple $(N, \{A_i\}_{i \in N}, \{r_i\}_{i \in N})$:
 - ▶ N is a finite set of players
 - ▶ A_i is a finite set of pure actions of player $i \in N$
 - ▶ r_i is the payoff function of player i : $r_i : A \mapsto \mathbb{R}$
 - ▶ where $A \equiv \times_{i \in N} A_i$ defines the set of action profiles
- ▶ Example: Prisoner's Dilemma

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- ▶ Example: Prisoner's Dilemma

		Player 2	
		C	D
Player 1	C	2, 2	-1, 4
	D	4, -1	0, 0

$$N = \{1, 2\},$$

$$A_1 = A_2 = \{C, D\},$$

$$r_1(C, C) = 2, r_1(C, D) = -1, r_1(D, C) = 4, \dots$$



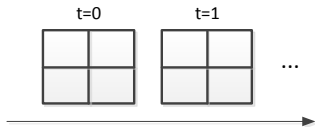
Repeated games

- ▶ In an infinitely repeated game, a certain stage-game is repeatedly played by the same set of players during an *a priori* unknown number of time-steps
- ▶ There is a probability of γ that the repeated game will continue after the current stage-game



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- ▶ The set of *histories up to time-step t* of the repeated game is given by $H^t \equiv \times_t A$
- ▶ The set of *all possible histories* is given by $H \equiv \bigcup_{t=0}^{\infty} H^t$ with $h \in H$ being a particular history
- ▶ A *mixed strategy of player i* is a mapping $\sigma_i : H \mapsto \Delta(A_i)$ with $\alpha_i \in \Delta(A_i)$ being a mixed action of player i



Nash equilibrium

- ▶ Let $\sigma_i \in \Sigma_i$ be a strategy of player i
- ▶ Let $\sigma \in \Sigma \equiv \times_i \Sigma_i$ be a strategy profile
- ▶ An *outcome path* is a possibly infinite sequence $\vec{a} \equiv (a^0, a^1, \dots)$ of action profiles
- ▶ The *discounted average payoff* of σ for player i is defined as

$$u_i^\gamma(\sigma) \equiv (1 - \gamma) \mathbb{E}_{\vec{a} \sim \sigma} \sum_{t=0}^{\infty} \gamma^t r_i(a^t),$$

- ▶ The discount factor can be seen as a patience of players: higher it is, more important are future payoffs
- ▶ A *Nash equilibrium* is defined as strategy profile $\sigma \equiv (\sigma_i, \sigma_{-i})$ such that for each player i and for every $\sigma'_i \in \Sigma_i$:

$$u_i^\gamma(\sigma) \geq u_i^\gamma(\sigma'_i, \sigma_{-i})$$



Subgame-perfect equilibrium

- ▶ A *subgame* is a repeated game which continues after a certain history
- ▶ For a pair (σ, h) , the *subgame strategy profile* induced by h is denoted as $\sigma|_h$
- ▶ A strategy profile σ is a *subgame-perfect equilibrium* (SPE) in a repeated game, if for all histories $h \in H$, the subgame strategy profile $\sigma|_h$ is a Nash equilibrium in the subgame

- ▶ Let be a stage-game:

		Player 2	
		<i>C</i>	<i>D</i>
Player 1	<i>C</i>	$r(C, C)$	$r(C, D)$
	<i>D</i>	$r(D, C)$	$r(D, D)$

- ▶ Given a strategy profile σ , after any history h^t , one can represent an (infinite) subgame as an *augmented stage-game*:

		Player 2	
		<i>C</i>	<i>D</i>
Player 1	<i>C</i>	$(1 - \gamma)r(C, C) + \gamma u^\gamma(\sigma _{h^t.(C,C)})$	$(1 - \gamma)r(C, D) + \gamma u^\gamma(\sigma _{h^t.(C,D)})$
	<i>D</i>	$(1 - \gamma)r(D, C) + \gamma u^\gamma(\sigma _{h^t.(D,C)})$	$(1 - \gamma)r(D, D) + \gamma u^\gamma(\sigma _{h^t.(D,D)})$

- ▶ The strategy profile σ is called subgame perfect equilibrium if it induces a Nash equilibrium in each augmented stage-game.



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Problem and Approach

- ▶ **Problem:** Given a discount factor γ and payoff functions of players, find the set of SPE entirely or partially
- ▶ Previous work includes:
 - ▶ All works on computing stage-game equilibria (ex: Lemke & Howson (1965), Porter et al. (2004))
 - ▶ Littman & Stone (2004): only for average payoff (i.e., $\gamma = 1$)
 - ▶ Judd et al. (2003): arbitrary γ but only pure action equilibria
- ▶ **Our approach:** dynamic programming over the set of equilibrium payoff profiles
 - ▶ Permits computing SPE for an arbitrary γ , including pure and mixed action equilibria
 - ▶ Based on two ideas: *self-generating sets* and *partitioning of hypercubes*

- ▶ Let $BR_i(\alpha)$ be a best response of player i in a stage-game to the mixed action profile $\alpha \equiv (\alpha_i, \alpha_{-i})$:

$$BR_i(\alpha) \equiv \max_{a_i \in A_i} r_i(a_i, \alpha_{-i}).$$

- ▶ We define the map B^γ on a set $W \subset \mathbb{R}^{|N|}$ as

$$B^\gamma(W) \equiv \bigcup_{(\alpha, w) \in \times_{i \in N} \Delta(A_i) \times W} (1 - \gamma)r(\alpha) + \gamma w,$$

- ▶ w is a *continuation promise* which verifies for all $i \in N$:

$$(1 - \gamma)r_i(\alpha) + \gamma w_i - (1 - \gamma)r_i(BR_i(\alpha), \alpha_{-i}) - \gamma \underline{w}_i \geq 0,$$

- ▶ $\underline{w}_i \equiv \inf_{w \in W} w_i$

- ▶ The largest fixed point of $B^\gamma(W)$ is the set of all SPE in the repeated game (Abreu, 1990)

- ▶ Recall the two self-generation equations:

$$B^\gamma(W) \equiv \bigcup_{(\alpha, w) \in \times_{i \in N} \Delta(A_i) \times W} (1 - \gamma)r(\alpha) + \gamma w \quad (1)$$

$$(1 - \gamma)r_i(\alpha) + \gamma w_i - (1 - \gamma)r_i(BR_i(\alpha), \alpha_{-i}) - \gamma \underline{w}_i \geq 0 \quad \forall i \quad (2)$$

- ▶ Equation (1) promises to player $i \in N$ a better payoff tomorrow to compensate a possible today's loss if player i follows a given strategy
- ▶ Equation (2) guarantees to player i a sufficient punishment imposed by the other players if player i deviates from the given strategy

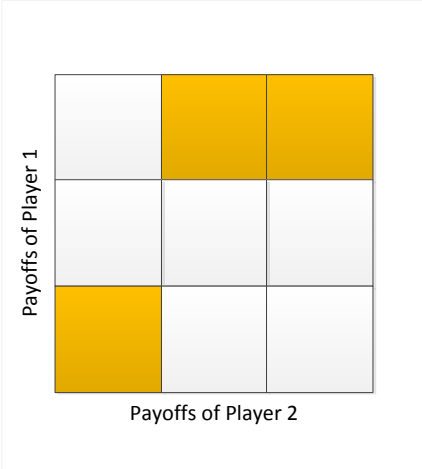


Updates by hypercubes

- ▶ Our algorithm starts with an initial approximation W of the set of SPE payoff profiles
- ▶ The set W , in turn, is represented by a union of disjoint hypercubes belonging to the set C
- ▶ Initially, the set C , contains only one hypercube that contains all possible payoff profiles
- ▶ Each iteration of the algorithm consists of verifying, for each hypercube $c \in C$, whether it has to be withdrawn



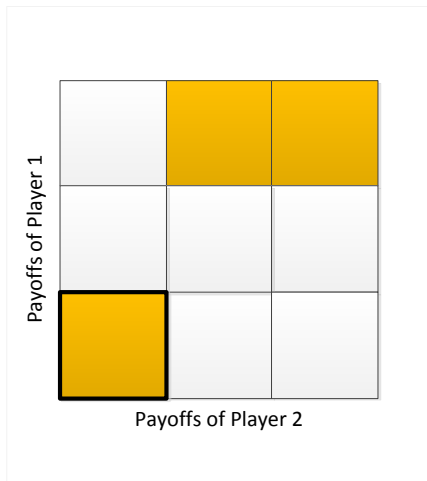
Updates by hypercubes: Example



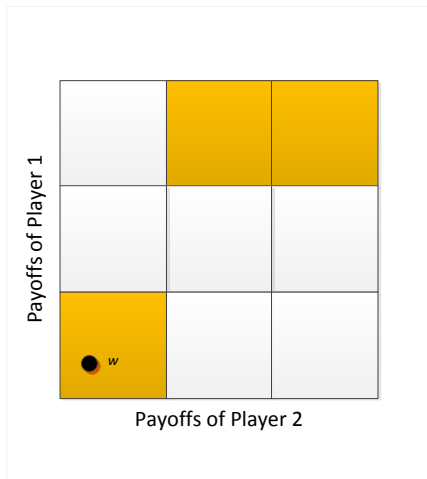
Payoffs of Player 1

Payoffs of Player 2

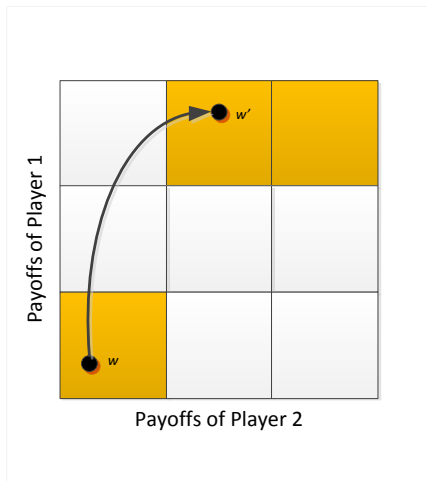
Updates by hypercubes: Example



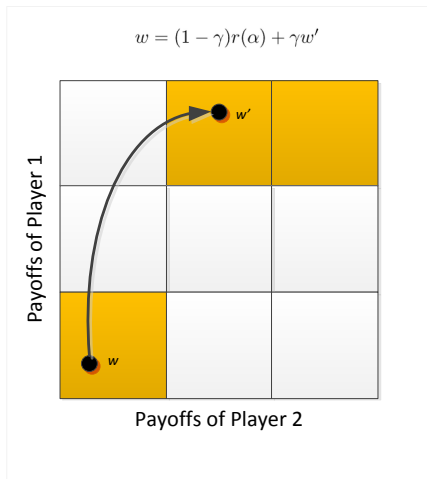
Updates by hypercubes: Example



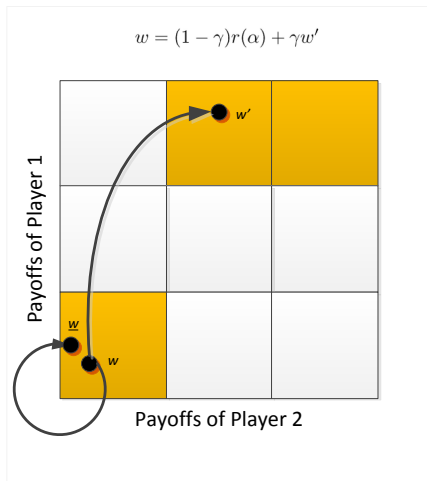
Updates by hypercubes: Example



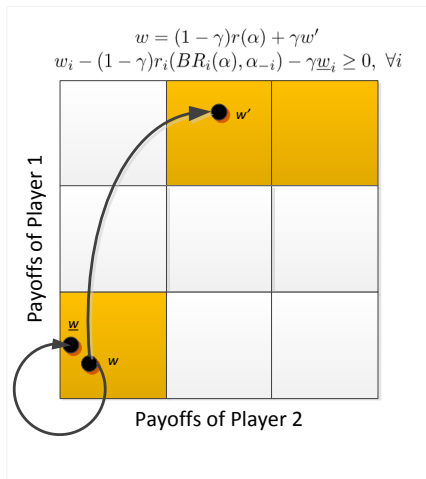
Updates by hypercubes: Example



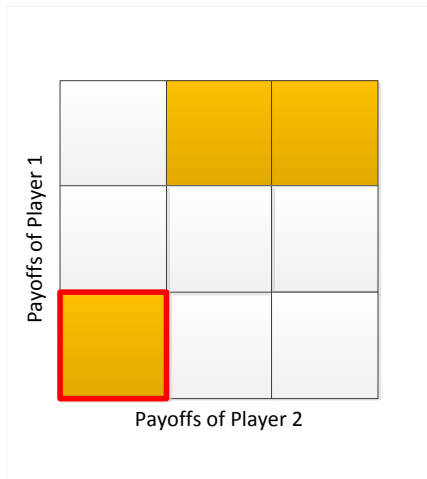
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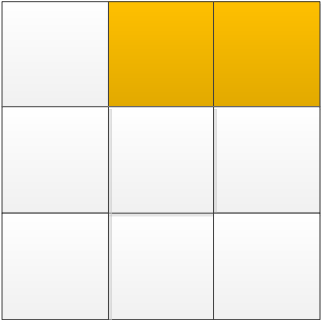


Updates by hypercubes: Example





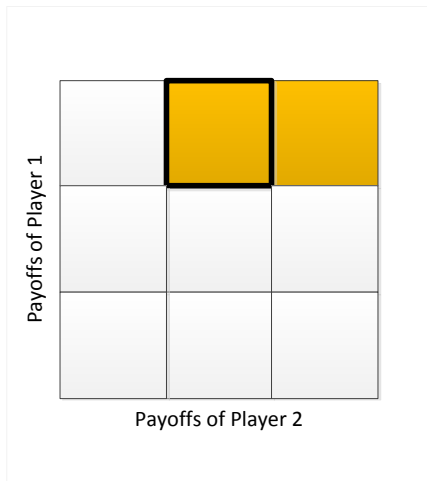
Updates by hypercubes: Example



Payoffs of Player 1

Payoffs of Player 2

Updates by hypercubes: Example

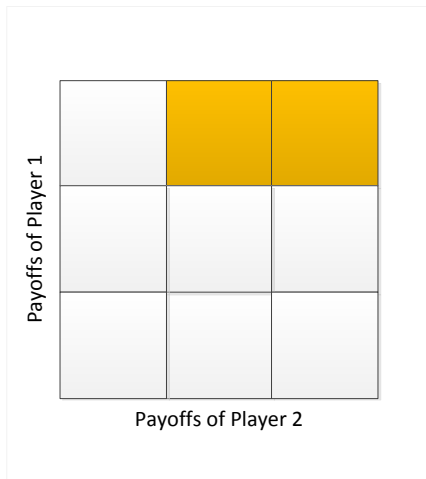




Partitioning the hypercubes

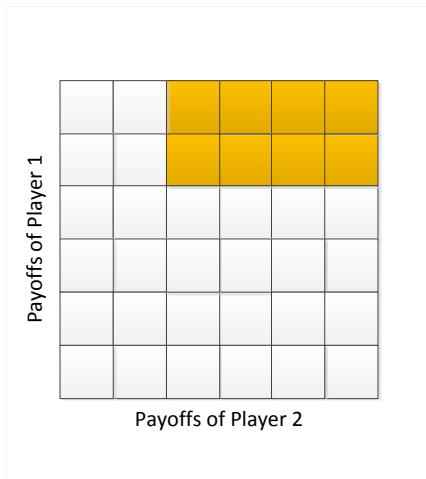
- ▶ If, after having tested all hypercubes in C , we haven't withdrawn any hypercube, we partition each remaining hypercube on a number of smaller hypercubes
 - ▶ We retest the remaining hypercubes the same way
 - ▶ This permits improving the precision of approximation of the set of equilibria
- ▶ The algorithm terminates when the required precision is achieved

Partitioning the hypercubes: Example



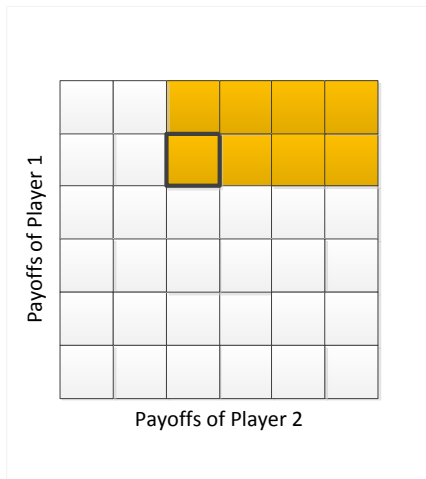


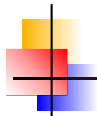
Partitioning the hypercubes: Example





Partitioning the hypercubes: Example





The Main Theorem

Theorem

For any repeated game, any discount factor γ and for any level of approximation, (i) Our algorithm terminates in finite time, (ii) the set of hypercubes C , at any moment, contains at least one hypercube, (iii) for any input $v \in W$, the algorithm returns a strategy profile (represented by a finite automaton) that satisfies the required approximation properties.



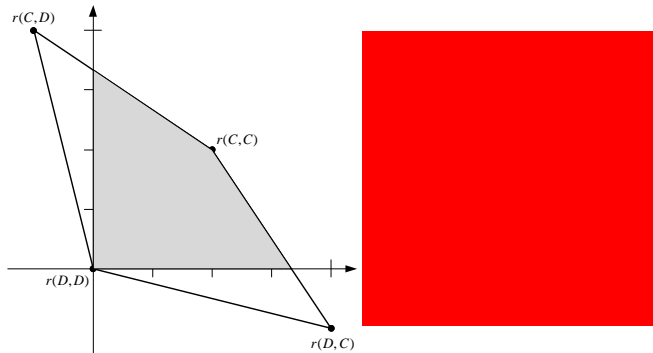
Example: *The Prisoner's Dilemma*

		Player 2	
		<i>C</i>	<i>D</i>
Player 1	<i>C</i>	2, 2	-1, 4
	<i>D</i>	4, -1	0, 0

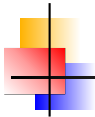
$$\gamma = 0.7$$



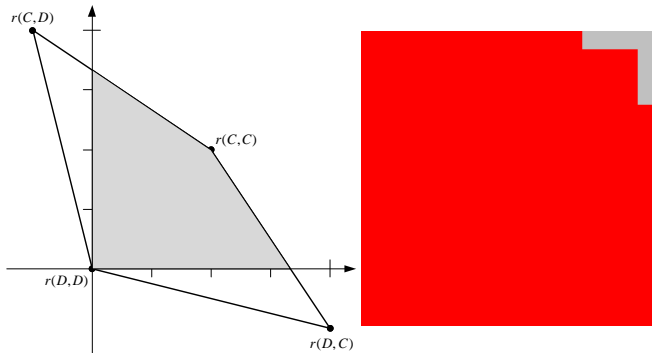
The Prisoner's Dilemma



Iteration 1



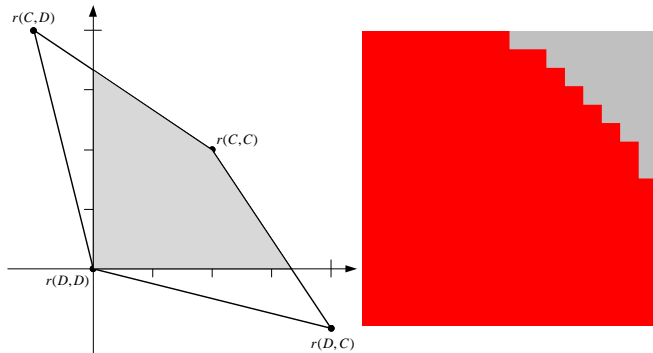
The Prisoner's Dilemma



Iteration 4



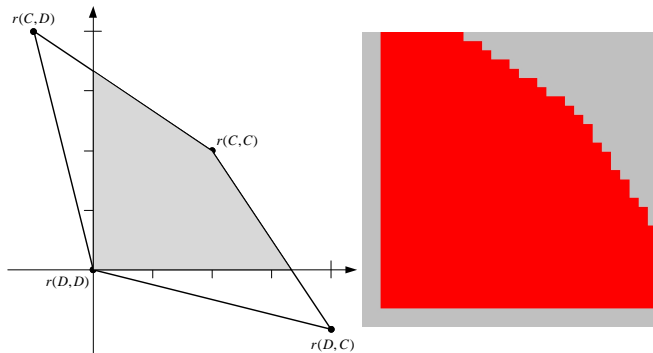
The Prisoner's Dilemma



Iteration 8



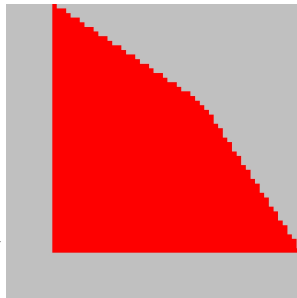
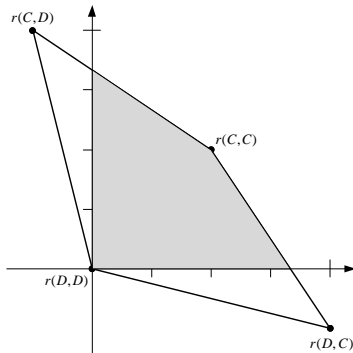
The Prisoner's Dilemma



Iteration 12



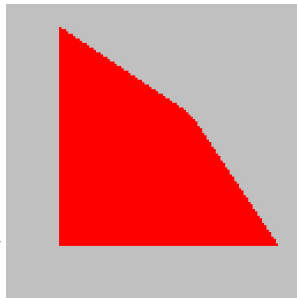
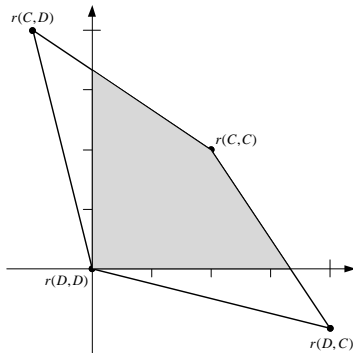
The Prisoner's Dilemma



Iteration 20



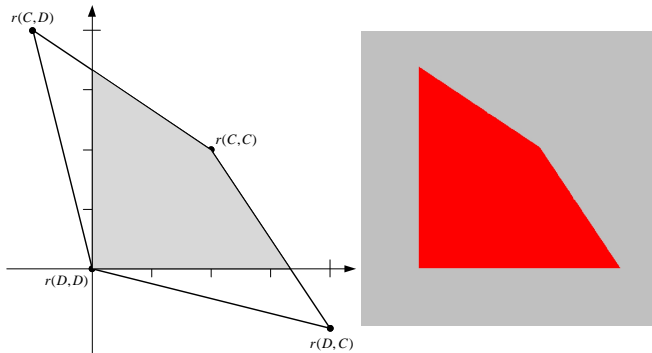
The Prisoner's Dilemma



Iteration 30



The Prisoner's Dilemma



Iteration 50



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Conclusion and Future Work

- ▶ We proposed an algorithmic approach for approximating the set of subgame-perfect equilibrium payoff profiles in repeated games
- ▶ Our algorithm is capable of computing a profile of player strategies that approximately induces any given SPE point
- ▶ Future work will aim at extending the proposed approach for solving more complex dynamic games such as Markov chain games and stochastic games



Thank you!



Another Example: Battle of the Sexes ($\gamma = 0.45$)

	<i>O</i>	<i>F</i>
<i>O</i>	1, 2	0, 0
<i>F</i>	0, 0	2, 1

Stage-game equilibrium payoff profiles:

- ▶ (1, 2)
- ▶ (2, 1)
- ▶ (2/3, 2/3)



Example: Repeated Battle of the Sexes





Example: Repeated Battle of the Sexes



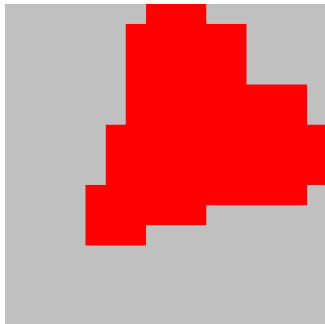


Example: Repeated Battle of the Sexes



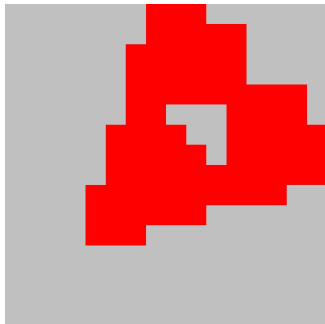


Example: Repeated Battle of the Sexes



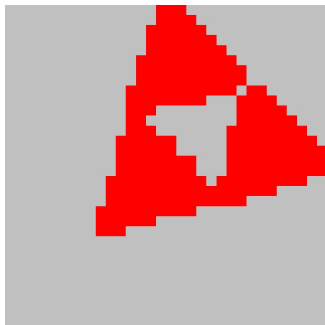


Example: Repeated Battle of the Sexes



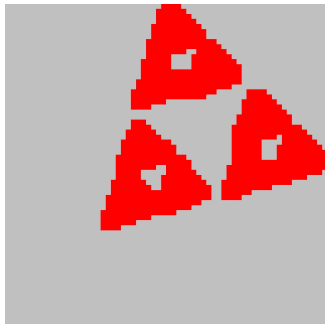


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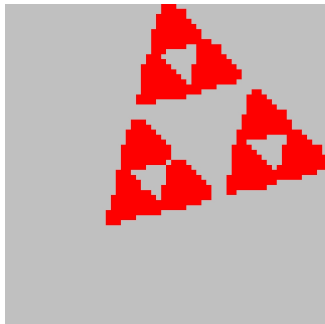


Example: Repeated Battle of the Sexes



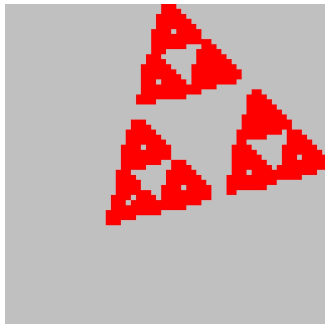


Example: Repeated Battle of the Sexes



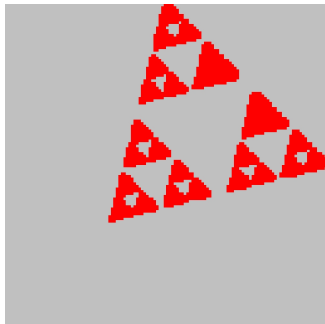


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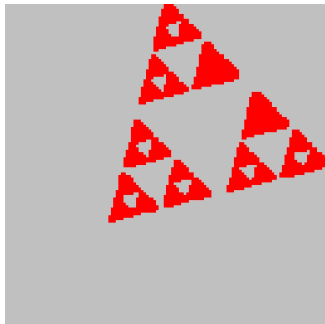


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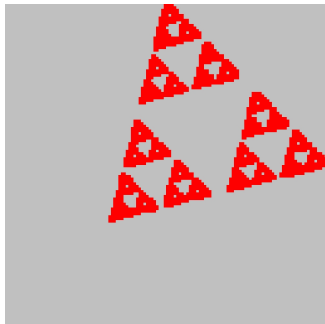


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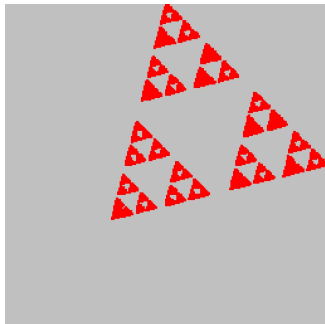


Example: Repeated Battle of the Sexes





Example: Repeated Battle of the Sexes





Automaton implementation

- ▶ Let $M \equiv (Q, q^0, f, \tau)$ be an *automaton implementation of a strategy profile σ* where
 - ▶ Q , set of automaton states with $q^0 \in Q$ being the initial state
 - ▶ $f \equiv (f_i)_{i \in N}$, where $f_i : Q \mapsto \Delta(A_i)$, la fonction de décision du joueur i
 - ▶ $\tau : Q \times A \mapsto Q$, une fonction de transition

Theorem (Kalai and Stanford, 1988)

Any SPE can be approximated by a finite automaton