



# Game THEORY

Nicolas Dudebout

Office DEPOT

Previously in *Game Theory*

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- ▶ decision makers:
  - ▶ choices
  - ▶ preferences

# Previously in *Game Theory*

- ▶ decision makers:
  - ▶ choices
  - ▶ preferences
- ▶ solution concepts:
  - ▶ best response
  - ▶ Nash equilibrium

Rock, paper, scissors

# Rock, paper, scissors

	$R$	$P$	$S$
$R$	0, 0	-1, 1	1, -1
$P$	1, -1	0, 0	-1, 1
$S$	-1, 1	1, -1	0, 0

# Learning in games

**Learning in games**

**Repeated games**



# Learning in games

# Best Response learning

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3. Observe the play

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1. Guess what the opponent(s) will play
2. Play a Best Response to that guess
3. Observe the play
4. Update the guess

# BR learning: Cournot dynamics

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Guess = last action played



# BR learning: Cournot dynamics

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	$C$	$D$
$C$	2, 2	-1, 3
$D$	3, -1	0, 0

# BR learning: Cournot dynamics

Guess = last action played

	$C$	$D$
$C$	2, 2	-1, 3
$D$	3, -1	0, 0

	$R$	$P$	$S$
$R$	0, 0	-1, 1	1, -1
$P$	1, -1	0, 0	-1, 1
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Guess = empirical distribution of play

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$R$	0, 0	-1, 1	1, -1
$P$	1, -1	0, 0	-1, 1
$S$	-1, 1	1, -1	0, 0

	$L$	$C$	$R$
$U$	0, 0	0, 1	1, 0
$M$	1, 0	0, 0	0, 1
$D$	0, 1	1, 0	0, 0

# Evolutionary learning

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Action set:  $A$

Utility function:  $u$



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Action set:  $A$

Utility function:  $u$

$$p \in \Delta(A), k \in A$$

$$\dot{p}_k = p_k (u(k, p) - u(p, p))$$

# Battle of the Sexes

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	<i>O</i>	<i>F</i>
<i>O</i>	3, 2	0, 0
<i>F</i>	0, 0	2, 3

# Correlated equilibrium (CE)

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$a^* \in A = \prod_i A_i$  is a NE:

$$\forall i, \forall a'_i, u_i(a_i^*, a_{-i}^*) \geq u_i(a'_i, a_{-i}^*)$$

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$\alpha \in \prod_i \Delta(A_i)$  is a CE:  $\forall i, \forall a_i, \forall a'_i,$

$$\sum_{a_{-i}} u_i(a_i, a_{-i}) \alpha(a) \geq \sum_{a_{-i}} u_i(a'_i, a_{-i}) \alpha(a)$$

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$\pi \in \Delta(A)$  is a CE:  $\forall i, \forall a_i, \forall a'_i,$

$$\sum_{a_{-i}} u_i(a_i, a_{-i})\pi(a) \geq \sum_{a_{-i}} u_i(a'_i, a_{-i})\pi(a)$$

# No regret learning



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$$u_i(k, a_{-i}) - u_i(j, a_{-i})$$

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$$u_i(k, a_{-i}) - u_i(j, a_{-i})$$

$$R_{jk}^i(t) = \sum_{\tau=0: a_i(\tau)=j}^t u_i(k, a_{-i}(\tau)) - u_i(j, a_{-i}(\tau))$$

# No regret learning

$$u_i(k, a_{-i}) - u_i(j, a_{-i})$$

$$R_{jk}^i(t) = \sum_{\tau=0:t}^{a_i(\tau)=j} u_i(k, a_{-i}(\tau)) - u_i(j, a_{-i}(\tau))$$

Regret matching converges to the correlated equilibria set.

# Learning in games

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- ▶ Best response

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- ▶ Replicator dynamics

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- ▶ Best response
- ▶ Replicator dynamics
- ▶ No regret

# Repeated games



# Markov Decision Process (MDP)

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state space  $X$

action space  $U$

transition  $P : X \times U \rightarrow \Delta(X)$

reward  $r : X \times U \rightarrow \mathbb{R}$

discount factor  $\delta \in [0, 1]$

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discount factor  $\delta \in [0, 1]$

$$U(x(\cdot), u(\cdot)) = \sum_{t=0}^{+\infty} \delta^t r(x(t), u(t))$$

# MDP (continued)

history  $\mathcal{H} \in \prod(X, U)$

policy  $\pi : \mathcal{H} \rightarrow \Delta(U)$

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policy  $\pi : \mathcal{H} \rightarrow \Delta(U)$

$$V^\pi(x_0) = \mathbb{E}_\pi [U(x(\cdot), u(\cdot))]$$

$$V(x_0) = \max_{\pi} V^\pi(x_0)$$

# Principle of Optimality

Bellman's equation:

$$V(x_0) = \max_{u_0} [r(x_0, u_0) + \delta V(P(x_0, u_0))]$$

# Dynamic Programming

Solving the MDP:



# Dynamic Programming

Solving the MDP:

- ▶ knowing  $P$ : value iteration

# Dynamic Programming

Solving the MDP:

- ▶ knowing  $P$ : value iteration
- ▶ not knowing  $P$ : online learning

# Repeated game

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Game  $(\mathcal{I}, \prod_i A_i, \prod_i u_i)$

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# Repeated game

Game  $(\mathcal{I}, \prod_i A_i, \prod_i u_i)$

Discount factor  $\delta$

$$U_i(a(\cdot)) = \sum_{t=0}^{+\infty} \delta^t u_i(a(t))$$

Strategy  $\sigma : \mathcal{H} \rightarrow \prod_i \Delta(A_i x)$

$$V_i(\sigma) = \mathbb{E}_\sigma [U_i(a(\cdot))]$$

# Nash equilibrium

Player  $i$ :

- ▶ choices  $\sigma_i$
- ▶ utility  $V_i$

# Nash equilibrium

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Nash equilibrium is not strong enough!  
(Explanation on the whiteboard  $\Rightarrow$ )



# Information structure

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- ▶ perfect
- ▶ imperfect

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- ▶ perfect
- ▶ imperfect
- ▶ public
- ▶ private (beliefs)

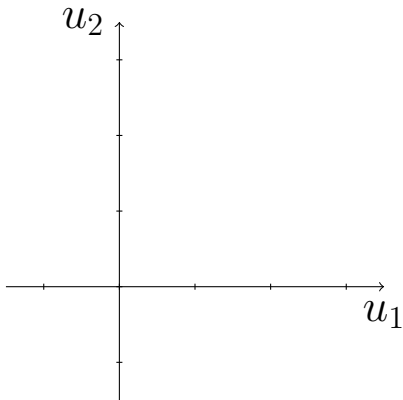
# Folk theorem

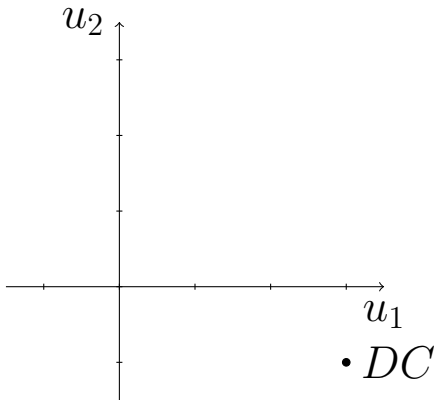
Any feasible, strictly individually rational payoff can be sustained by a sequentially rational equilibrium.

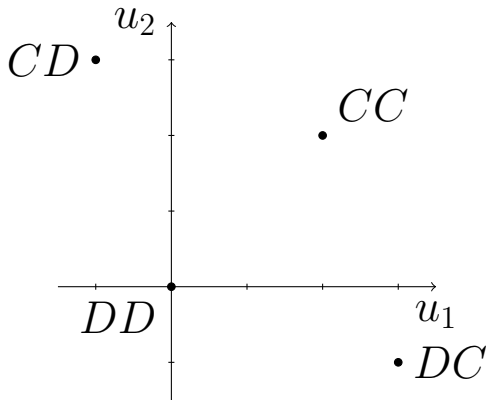
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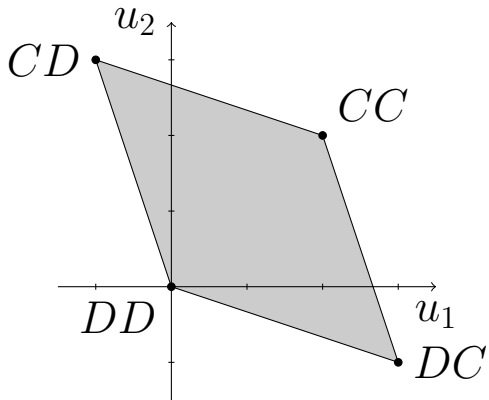
Holy grail for repeated games.

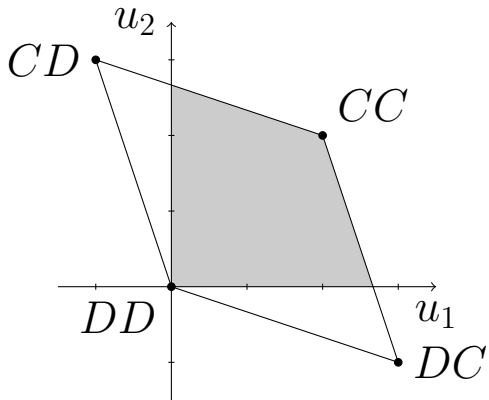












**Research**

# Weakly belief-free equilibria

Characterization of repeated games with correlated equilibria.

# Repeated games

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- ▶ Folk theorem



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**Questions,  
Comments**