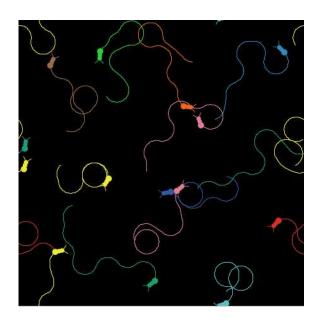
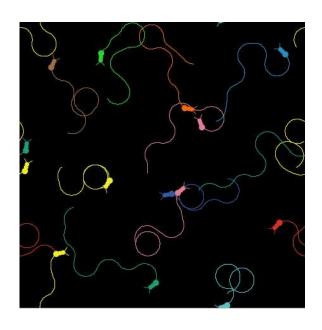
Multi-agent learning

Fictitious Play

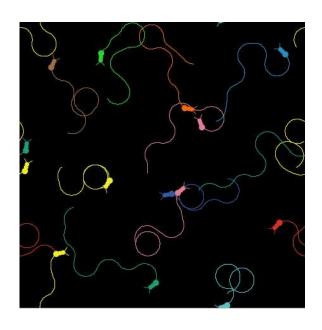
Gerard Vreeswijk, Intelligent Software Systems, Computer Science Department, Faculty of Sciences, Utrecht University, The Netherlands.

Wednesday 19th May, 2021

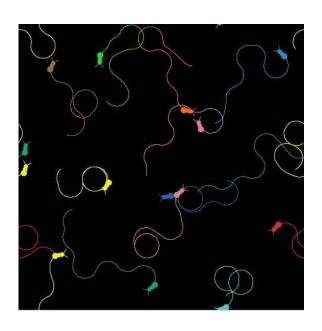




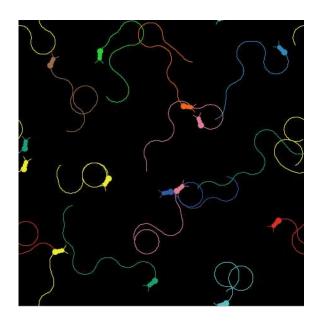
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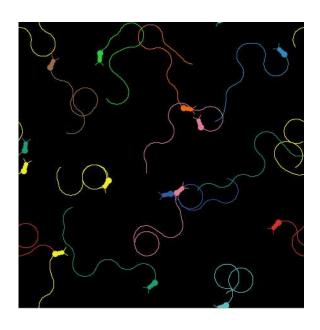
■ One of the most important, if not the most important, representative of a follower strategy.



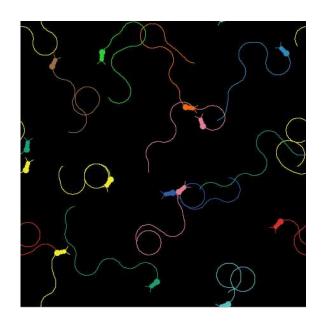
- One of the most important, if not the most important, representative of a follower strategy.
- Rather than considering your own payoffs, monitor the behaviour of your opponent(s).



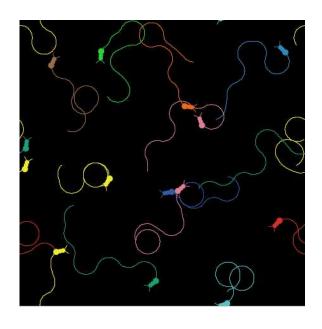
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- Brown (1951): explanation for Nash equilibrium play. In terms of current use, the name actually is a bit of a misnomer, since play actually occurs (Berger, 2005).

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Part I. Best reply strategy

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1. Pure fictitious play.

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Part II. Extensions and approximations of fictitious play

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Shoham et al. (2009): Multi-agent Systems. Ch. 7: "Learning and Teaching". H. Young (2004): Strategic Learning and it Limits, Oxford UP. D. Fudenberg and D.K. Levine (1998), The Theory of Learning in Games, MIT Press.

Part I: Pure fictitious play

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* = chosen randomly.

Round A's action B's action A's beliefs B's beliefs

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Round	A's action	B's action	A's beliefs	B's beliefs
0.				

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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0,0)	(0,0)

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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0,0)	(0,0)
1.	Γ_*	R*		

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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0,0)	(0,0)
1.	L^*	R*	(0,1)	(1,0)
2.	R	L		

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0.			(0,0)	(0,0)
1.	L*	R*	(0,1)	(1,0)
2.	R	L	(1,1)	(1,1)
3.	L*	R*	, ,	` ,

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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0,0)	(0,0)
1.	L*	R*	(0,1)	(1,0)
2.	R	L	(1, 1)	(1, 1)
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0.			(0,0)	(0,0)
1.	L*	R*	(0,1)	(1,0)
2.	R	L	(1,1)	(1, 1)
3.	L^*	R*	(1,2)	(2,1)
4.	R	L	(2,2)	(2,2)
5.	R*	R*	, ,	, ,

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4.	R	L	(2, 2)	(2,2)
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3.	L*	R*	(1, 2)	(2, 1)
4.	R	L	(2,2)	(2,2)
5.	R*	R*	(2,3)	(2,3)
6.	R	R	, ,	,

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2.	R	L	(1,1)	(1, 1)
3.	L^*	R*	(1,2)	(2, 1)
4.	R	L	(2,2)	(2,2)
5.	R*	R*	(2,3)	(2,3)
6.	R	R	(2,4)	(2,4)

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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0,0)	(0,0)
1.	L*	R*	(0,1)	(1,0)
2.	R	L	(1, 1)	(1,1)
3.	L*	R*	(1,2)	(2, 1)
4.	R	L	(2,2)	(2,2)
5.	R*	R*	(2,3)	(2,3)
6.	R	R	(2,4)	(2,4)
7.	R	R	, ,	` '

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4.	R	L	(2,2)	(2,2)
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4.	R	L	(2,2)	(2,2)
5.	R*	R*	(2,3)	(2,3)
6.	R	R	(2,4)	(2,4)
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:	: :	: :	• •	• •



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There are many more equilibrium types (semi-strict, locally stable, ϵ -equilibrium, correlated, coarse correlated, ...). Which equilibrium types imply which other equilibrium types is an interesting question.

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Proof. Suppose the pure action profile a is a strict Nash equilibrium. Suppose a is played at round t. Because a is strict, a_i is the unique best response to a_{-i} , for each i.

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Summary of the two theorems:

Strict and pure Nash \Rightarrow Steady state \Rightarrow Pure Nash.

Theorem. Suppose a pure strategy profile is a <u>strict</u> Nash equilibrium of a stage game. Then it must be a steady state of fictitious play.

Notice the use of terminology:

- "pure strategy profile" for Nash equilibria.
- "steady state" for fictitious play.

Proof. Suppose the pure action profile a is a strict Nash equilibrium. Suppose a is played at round t. Because a is strict, a_i is the unique best response to a_{-i} , for each i. Therefore, the action profile a will be played in round t+1 again.

Summary of the two theorems:

Strict and pure Nash \Rightarrow Steady state \Rightarrow Pure Nash.

But what if all equilibria are mixed?

Example. Matching Pennies.

Example. Matching Pennies. This is a zero-sum game.

Example. Matching Pennies. This is a zero-sum game. A's goal is to have pennies matched. B's goal is the opposite.

Round A's action B's action A's beliefs B's beliefs

Round	A's action	B's action	A's beliefs	B's beliefs
0.				

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T		

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	${ m T}$	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н		,

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н		,

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н		

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)
5.	Н	Н	,	. ,

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)
5.	Н	Н	(5.5, 3.0)	(4.0, 4.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)
5.	Н	Н	(5.5, 3.0)	(4.0, 4.5)
6.	Н	Н	,	,

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	H	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)
5.	Н	H	(5.5, 3.0)	(4.0, 4.5)
6.	Н	Н	(6.5, 3.0)	(5.0, 4.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)
5.	Н	Н	(5.5, 3.0)	(4.0, 4.5)
6.	Н	Н	(6.5, 3.0)	(5.0, 4.5)
7.	Н	T	,	` '

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)
5.	Н	Н	(5.5, 3.0)	(4.0, 4.5)
6.	Н	Н	(6.5, 3.0)	(5.0, 4.5)
7.	H	T	(6.5, 4.0)	(6.0, 4.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)
5.	Н	Н	(5.5, 3.0)	(4.0, 4.5)
6.	Н	Н	(6.5, 3.0)	(5.0, 4.5)
7.	Н	T	(6.5, 4.0)	(6.0, 4.5)
8.	Н	T		

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)
5.	Н	Н	(5.5, 3.0)	(4.0, 4.5)
6.	Н	Н	(6.5, 3.0)	(5.0, 4.5)
7.	Н	T	(6.5, 4.0)	(6.0, 4.5)
8.	H	T	(6.5, 5.0)	(7.0, 4.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(1.5, 2.0)	(2.0, 1.5)
1.	T	T	(1.5, 3.0)	(2.0, 2.5)
2.	T	Н	(2.5, 3.0)	(2.0, 3.5)
3.	T	Н	(3.5, 3.0)	(2.0, 4.5)
4.	Н	Н	(4.5, 3.0)	(3.0, 4.5)
5.	Н	Н	(5.5, 3.0)	(4.0, 4.5)
6.	Н	Н	(6.5, 3.0)	(5.0, 4.5)
7.	Н	T	(6.5, 4.0)	(6.0, 4.5)
8.	Н	T	(6.5, 5.0)	(7.0, 4.5)
•	•	•	:	:

Frequencies of fictitious play

```
☑ 4 $ ... 3D
                    setup
     clear-drawing
                                            matrix: [[[1 -1] [-1 1]] [[-1 1] [1 -1]]]
                                            action: [1 1]
                                            frequencies: [[0 1] [0 1]]
           step
                     go
                                            expected-rewards: [[-1 1] [1 -1]]
game-type
                                            matrix: [[[1 -1] [-1 1]] [[-1 1] [1 -1]]]
matching-pennies
                        V
                                            action: [1 0]
                                            frequencies: [[0 2] [1 1]]
                                            expected-rewards: [[0 0] [1 -1]]
   nr-of-actions
                                            matrix: [[[1 -1] [-1 1]] [[-1 1] [1 -1]]]
                                            action: [1 0]
   epsilon
                      0.10
                                            frequencies: [[0 3] [2 1]]
                                            expected-rewards: [[0.3333333333333 -0.33333333333333
   initial-cumulative
                                            matrix: [[[1 -1] [-1 1]] [[-1 1] [1 -1]]]
                                            action: [0 0]
   initial-geometric
                       50
                                            frequencies: [[1 3] [3 1]]
                                            expected-rewards: [[0.5 -0.5] [0.5 -0.5]]
   learning-rate
                      0.20
                                            matrix: [[[1 -1] [-1 1]] [[-1 1] [1 -1]]]
                                            action: [0 0]
                                            frequencies: [[2 3] [4 1]]
   max-payoff
                      100
                                            matrix: [[[1 -1] [-1 1]] [[-1 1] [1 -1]]]
   penalty
                       -15
                                            action: [0 0]
                                            frequencies: [[3 3] [5 1]]
   lambda
                     0.100
                                            expected-rewards: [[0.66666666666667 -0.666666666666666
```

Theorem. If the empirical distribution of strategies converges in fictitious play, then it converges to a Nash equilibrium.

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Proof. Let *i* be arbitrary. If the empirical distribution converges to q, then i's opponent model converges to q^{-i} . By definition of fictitious play, $q^i \in BR(q^{-i})$. Because of convergence, all such (mixed) best replies converge along.

Theorem. If the empirical distribution of strategies converges in fictitious play, then it converges to a Nash equilibrium.

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Remarks:

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Remarks:

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Remarks:

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Remarks:

- 1. The q^i may be mixed.
- 2. It actually suffices that the q^{-i} converge asymptotically to the actual distribution (Fudenberg & Levine, 1998).
- 3. If the empirical distributions converge (hence, converge to a Nash equilibrium), the actually played responses per stage need not be Nash equilibria of the stage game.

Repeated Coordination Game. Players receive payoff 1 iff they coordinate, else 0.

Round A's action B's action A's beliefs B's beliefs

Round	A's action	B's action	A's beliefs	B's beliefs
0.				

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A		

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)
2.	A	В		

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)
2.	A	В	(1.5, 2.0)	(2.0, 1.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)
2.	A	В	(1.5, 2.0)	(2.0, 1.5)
3.	В	A	` '	

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)
2.	A	В	(1.5, 2.0)	(2.0, 1.5)
3.	В	A	(2.5, 2.0)	(2.0, 2.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)
2.	A	В	(1.5, 2.0)	(2.0, 1.5)
3.	В	A	(2.5, 2.0)	(2.0, 2.5)
4.	A	В	,	

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)
2.	A	В	(1.5, 2.0)	(2.0, 1.5)
3.	В	A	(2.5, 2.0)	(2.0, 2.5)
4.	A	В	(2.5, 3.0)	(3.0, 2.5)

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)
2.	A	В	(1.5, 2.0)	(2.0, 1.5)
3.	В	A	(2.5, 2.0)	(2.0, 2.5)
4.	A	В	(2.5, 3.0)	(3.0, 2.5)
•	• •	•	: :	:

Repeated Coordination Game. Players receive payoff 1 iff they coordinate, else 0.

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)
2.	A	В	(1.5, 2.0)	(2.0, 1.5)
3.	В	A	(2.5, 2.0)	(2.0, 2.5)
4.	A	В	(2.5, 3.0)	(3.0, 2.5)
•	:	• •	: :	: :

■ This game possesses three equilibria, viz. (0,0), (0.5,0.5), and (1,1), with expected payoffs 1, 0.5, and 1, respectively.

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.5, 1.0)	(1.0, 0.5)
1.	В	A	(1.5, 1.0)	(1.0, 1.5)
2.	A	В	(1.5, 2.0)	(2.0, 1.5)
3.	В	A	(2.5, 2.0)	(2.0, 2.5)
4.	A	В	(2.5, 3.0)	(3.0, 2.5)
• •	: :	: :	: :	: :

- This game possesses three equilibria, viz. (0,0), (0.5,0.5), and (1,1), with expected payoffs 1, 0.5, and 1, respectively.
- Empirical distribution of play converges to (0.5, 0.5),—with payoff 0, rather than 0.5.

Rock-paper-scissors. Winner receives payoff 1, else 0.

■ Rock-paper-scissors with these payoffs is known as Shapley's game.

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Round A's action B's action A's beliefs B's beliefs

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Round	A's action	B's action	A's beliefs	B's beliefs
0.				

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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)

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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)
1.	Rock	Scissors		

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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)
1.	Rock	Scissors	(0.0, 0.0, 1.5)	(1.0, 0.5, 0.0)

- Rock-paper-scissors with these payoffs is known as Shapley's game.
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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)
1.	Rock	Scissors	(0.0, 0.0, 1.5)	(1.0, 0.5, 0.0)
2.	Rock	Paper		

- Rock-paper-scissors with these payoffs is known as Shapley's game.
- Shapley's game possesses one equilibrium, viz. (1/3, 1/3, 1/3), with expected payoff 1/3.

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)
1.	Rock	Scissors	(0.0, 0.0, 1.5)	(1.0, 0.5, 0.0)
2.	Rock	Paper	(0.0, 1.0, 1.5)	(2.0, 0.5, 0.0)

- Rock-paper-scissors with these payoffs is known as Shapley's game.
- Shapley's game possesses one equilibrium, viz. (1/3, 1/3, 1/3), with expected payoff 1/3.

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)
1.	Rock	Scissors	(0.0, 0.0, 1.5)	(1.0, 0.5, 0.0)
2.	Rock	Paper	(0.0, 1.0, 1.5)	(2.0, 0.5, 0.0)
3.	Rock	Paper	,	,

- Rock-paper-scissors with these payoffs is known as Shapley's game.
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Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)
1.	Rock	Scissors	(0.0, 0.0, 1.5)	(1.0, 0.5, 0.0)
2.	Rock	Paper	(0.0, 1.0, 1.5)	(2.0, 0.5, 0.0)
3.	Rock	Paper	(0.0, 2.0, 1.5)	(3.0, 0.5, 0.0)

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- Shapley's game possesses one equilibrium, viz. (1/3, 1/3, 1/3), with expected payoff 1/3.

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)
1.	Rock	Scissors	(0.0, 0.0, 1.5)	(1.0, 0.5, 0.0)
2.	Rock	Paper	(0.0, 1.0, 1.5)	(2.0, 0.5, 0.0)
3.	Rock	Paper	(0.0, 2.0, 1.5)	(3.0, 0.5, 0.0)
4.	Scissors	Paper		

- Rock-paper-scissors with these payoffs is known as Shapley's game.
- Shapley's game possesses one equilibrium, viz. (1/3, 1/3, 1/3), with expected payoff 1/3.

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)
1.	Rock	Scissors	(0.0, 0.0, 1.5)	(1.0, 0.5, 0.0)
2.	Rock	Paper	(0.0, 1.0, 1.5)	(2.0, 0.5, 0.0)
3.	Rock	Paper	(0.0, 2.0, 1.5)	(3.0, 0.5, 0.0)
4.	Scissors	Paper	(0.0, 3.0, 1.5)	(3.0, 0.5, 1.0)

Empirical distr. of play does not need to converge

Rock-paper-scissors. Winner receives payoff 1, else 0.

- Rock-paper-scissors with these payoffs is known as Shapley's game.
- Shapley's game possesses one equilibrium, viz. (1/3, 1/3, 1/3), with expected payoff 1/3.

Round	A's action	B's action	A's beliefs	B's beliefs
0.			(0.0, 0.0, 0.5)	(0.0, 0.5, 0.0)
1.	Rock	Scissors	(0.0, 0.0, 1.5)	(1.0, 0.5, 0.0)
2.	Rock	Paper	(0.0, 1.0, 1.5)	(2.0, 0.5, 0.0)
3.	Rock	Paper	(0.0, 2.0, 1.5)	(3.0, 0.5, 0.0)
4.	Scissors	Paper	(0.0, 3.0, 1.5)	(3.0, 0.5, 1.0)
5.	Scissors	Paper		

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1.	Rock	Scissors	(0.0, 0.0, 1.5)	(1.0, 0.5, 0.0)
2.	Rock	Paper	(0.0, 1.0, 1.5)	(2.0, 0.5, 0.0)
3.	Rock	Paper	(0.0, 2.0, 1.5)	(3.0, 0.5, 0.0)
4.	Scissors	Paper	(0.0, 3.0, 1.5)	(3.0, 0.5, 1.0)
5.	Scissors	Paper	(0.0, 4.0, 1.5)	(3.0, 0.5, 2.0)

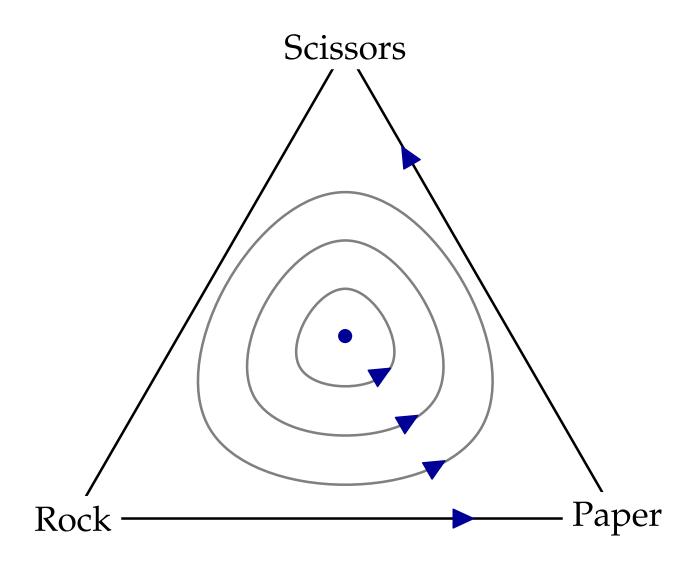
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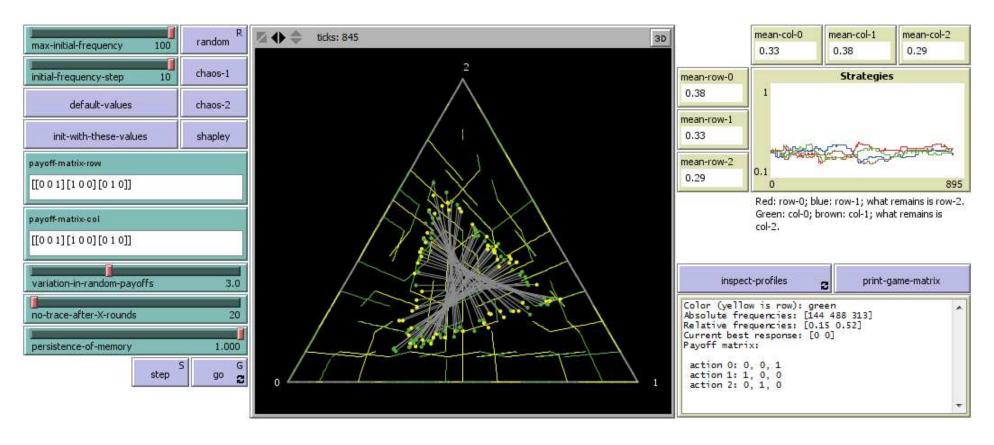
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2.	Rock	Paper	(0.0, 1.0, 1.5)	(2.0, 0.5, 0.0)
3.	Rock	Paper	(0.0, 2.0, 1.5)	(3.0, 0.5, 0.0)
4.	Scissors	Paper	(0.0, 3.0, 1.5)	(3.0, 0.5, 1.0)
5.	Scissors	Paper	(0.0, 4.0, 1.5)	(3.0, 0.5, 2.0)
•	•	•	• •	• •

Repeated Shapley game: phase diagram

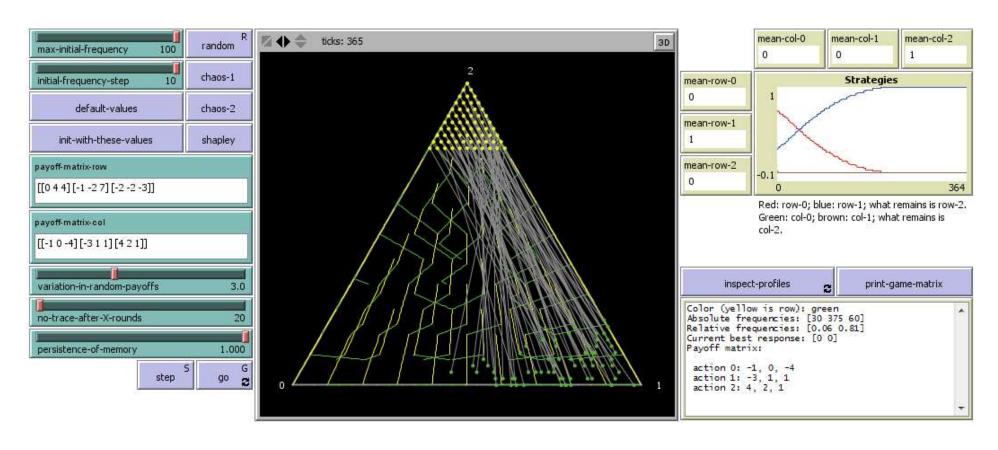


FP on Shapley's game; strategy profiles in a simplex



There are many player couples. Each couple is connected by a gray line. Yellow is row; green is column. Player location is determined by the mixed strategy it projects on its opponent (i.e., normalised action count of its opponent). Each player starts with a biased action count. For example, with [100,0,0] (lower left) or [0,100,0] (lower right) or [33,33,33] (center). Initial action counts of player pairs are unrelated.

FP on a 3x3 game; strategy profiles in a simplex



There are many player couples. Each couple is connected by a gray line. Yellow is row; green is column. Player location is determined by the mixed strategy it projects on its opponent (i.e., normalised action count of its opponent). Each player starts with a biased action count. For example, with [100,0,0] (lower left) or [0,100,0] (lower right) or [33,33,33] (center). Initial action counts of player pairs are unrelated.

Part II: Extensions and approximations of fictitious play



Author: Gerard Vreeswijk. Slides last modified on May 19th, 2021 at 16:50

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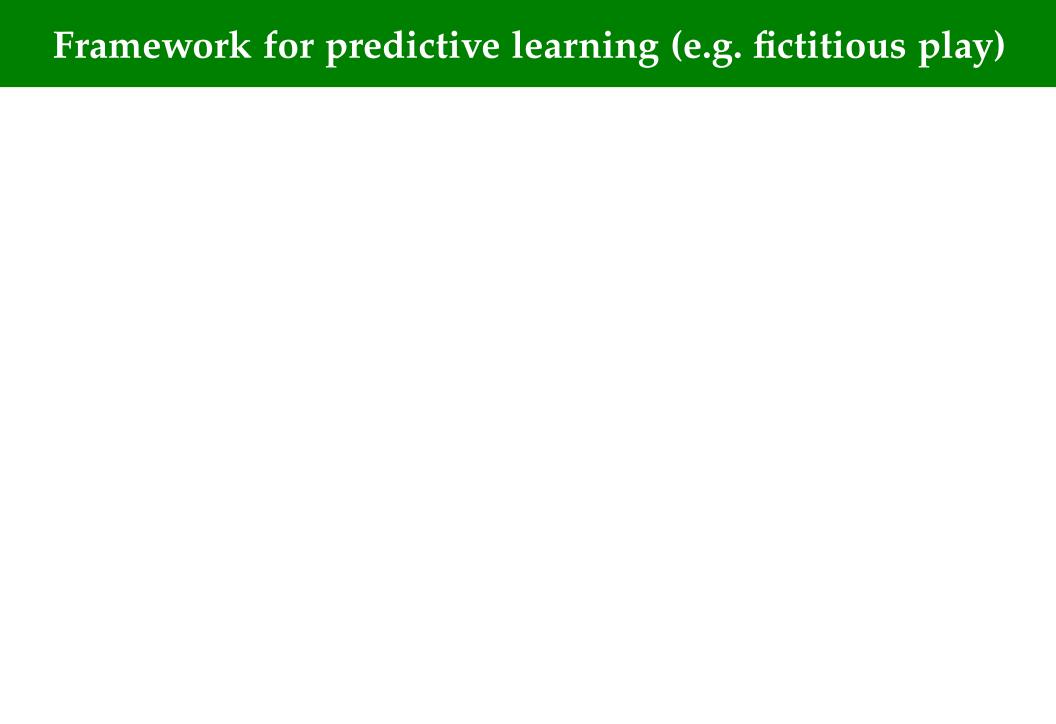
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 - **Perturbed throughout**, with small random shocks.
 - Randomly, and **proportional to expected payoff**.

Jordan's framework for FP



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A predictive learning rule for player *i* is the combination of a forecasting rule and a response rule.

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■ This framework can be attributed to J.S. Jordan (1993).

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Forecasting and response rules for fictitious play

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Elaborations of this idea:

b) Through soft max (a.k.a. mixed logit):

$$q^{i}(x_{i} \mid p^{-i}) =_{Def} \frac{e^{u_{i}(x_{i},p^{-i})/\gamma}}{\sum_{x'_{i} \in X_{i}} e^{u_{i}(x'_{i},p^{-i})/\gamma}}.$$

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a) Strictly proportional:

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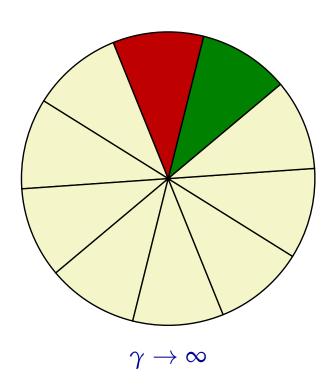
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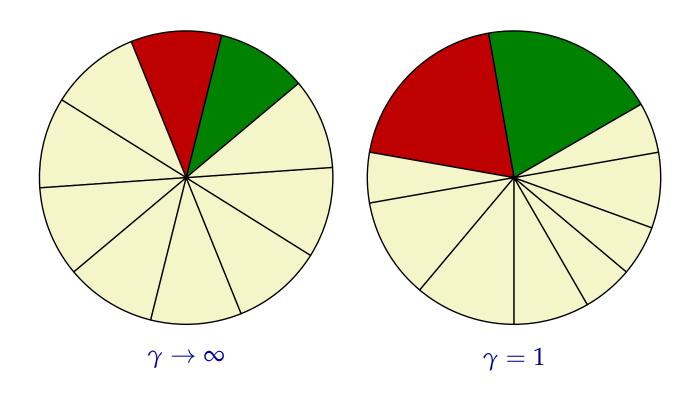
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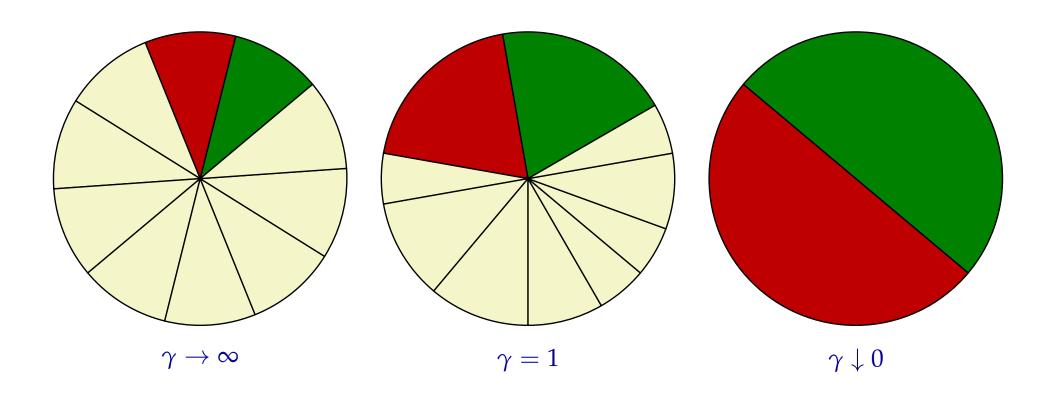
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Fudenberg & Levine, 1995. "Consistency and cautious fictitious play," *Journal of Economic Dynamics and Control*, Vol. **19** (5-7), pp. 1065-1089.

Hart & Mas-Colell, 2001. "A General Class of Adaptive Strategies," *Journal of Economic Theory*, Vol. **98**(1), pp. 26-54.

Definition. Let X be action profiles, and $q \in \Delta(X)$. Then q is a coarse correlated equilibrium (CCE) if no one wants to opt out prior to a realisation of q in the form of an action profile.

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Definition. Let X be action profiles, and $q \in \Delta(X)$. Then q is a coarse correlated equilibrium (CCE) if no one wants to opt out prior to a realisation of q in the form of an action profile.

In a coarse correlated ϵ -equilibrium (ϵ -CCE), no player wants to opt out to gain more in expected utility than ϵ .

Theorem (Fudenberg & Levine, 1995). Let G be a finite game and let $\epsilon > 0$. If all players use smoothed fictitious play with sufficiently small smoothing parameters γ_i , then empirical play converges to the set of coarse correlated ϵ -equilibria a.s.

Summary of the two theorems:

- Smoothed fictitious play limits regret.
- Smoothed fictitious play converges to ϵ -CCE.

But there is another l.a. with **no** regret and convergence to **zero-**CCE!

Exponentiated regret matching



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$$q_j^{i(t+1)} \propto [\bar{r}_j^{it}]_+^{\mathbf{a}}$$

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A theorem on exponentiated regret matching (Mas-Colell *et al.*, 2001) ensures that individual players have no-regret with probability one, and the empirical distribution of play converges to the set of coarse correlated equilibria (PY, p. 37 for RM, p. 60 for ERM).

FP

FP vs. Smoothed FP

FP vs. Smoothed FP vs. Exponentiated regret matching



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Hart, S., and Mas-Colell, A. (2000). "A simple adaptive procedure leading to correlated equilibrium". *Econometrica*, **68**, pp. 1127-1150.



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FP Smoothed FP Exponentiated RM

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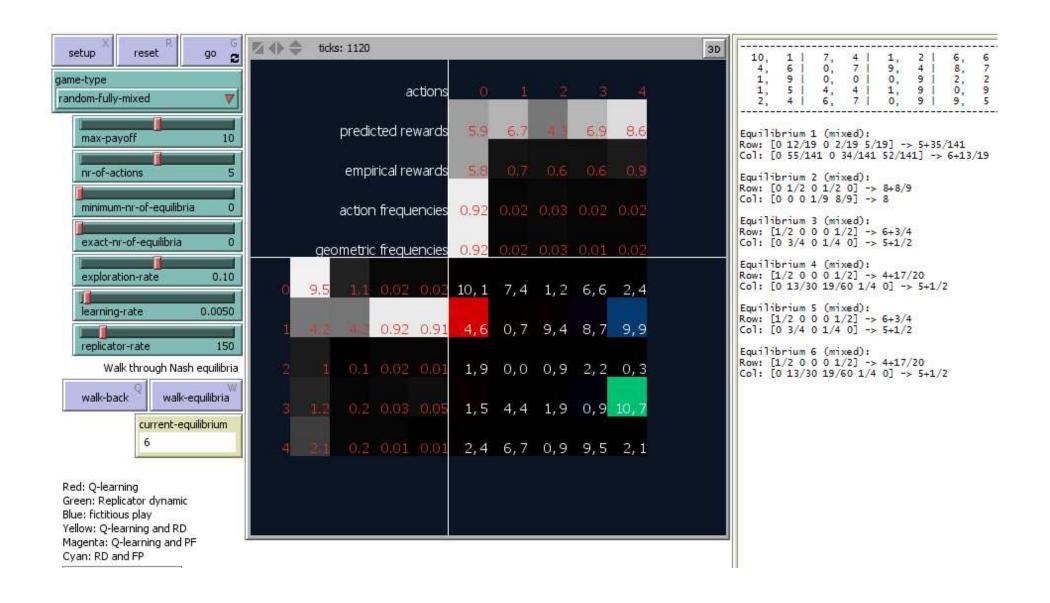
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Collective convergence to coarse correlated equilibria	_	Within $\epsilon > 0$, almost always (PY, p. 83)	Exact, almost always (PY, p. 60)

Fictitious play compared to other algorithms



Part III: Finite memory and inertia



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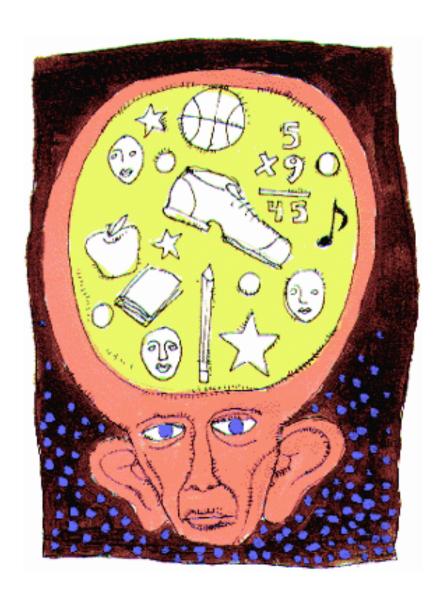
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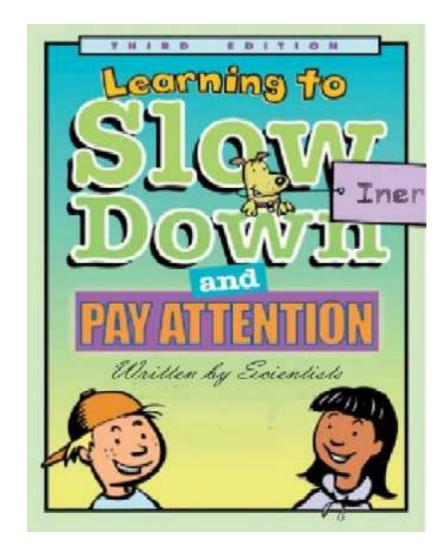
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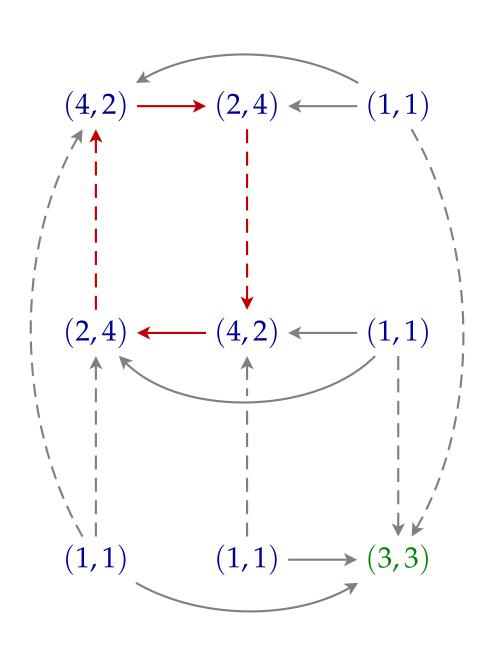
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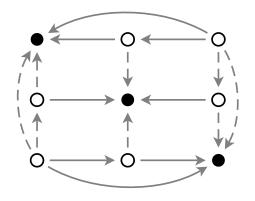
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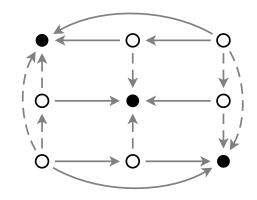


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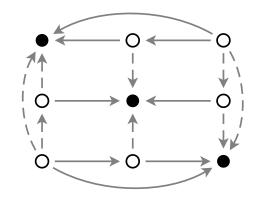
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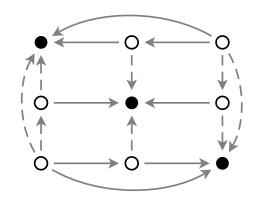
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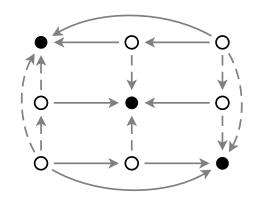
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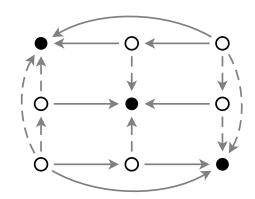


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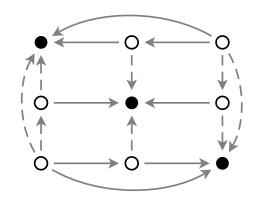
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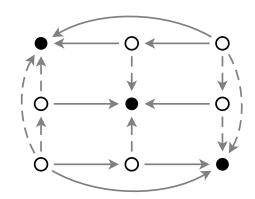
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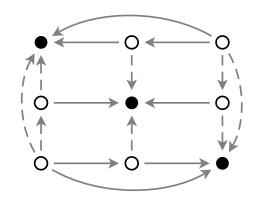
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Examples of weakly acyclic games

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6. It can be shown that, due to weak acyclicity, inertia, and (4), the process eventually lands in an absorbing state which, due to (5), is a repeated pure Nash equilibrium. □



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Hence,

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Infinitely many disjoint histories of length \mathbf{m} occur, hence infinitely many *independent* events "homogeneous at $t + \mathbf{m}$ " occur. Apply the (second) Borel-Cantelli lemma: if $\{E_n\}_n$ are independent events, and $\sum_{n=1}^{\infty} \Pr(E_n)$ is unbounded, then $\Pr($ an infinite number of E_n 's occur)=1. \square

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Final claim: probability to reach a sink from $Z^* > 0$

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Since Z^* is encountered infinitely often, the result follows.





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 ii) can be brought arbitrary close to fictitious play.
- In weakly acyclic *n*-person games, every better-reply process with finite memory and inertia converges to a pure Nash equilibrium.

- Like fictitious play, players model (or assess) each other through mixed strategies.
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Exam problems



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Problem. *A*: policy holder is accident prone; A_1 : policy holder will have an accident within one year. Suppose $\Pr\{A_1|A\} = 0.4$, $\Pr\{A_1|\overline{A}\} = 0.2$ and $\Pr\{A\} = 0.3$. Compute $\Pr\{A_1\}$.

Answer.

If *E* and *F* are events, then

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$$= Pr\{EF\} + Pr\{E\overline{F}\}$$

$$= Pr\{E|F\}Pr\{F\} + Pr\{E|\overline{F}\}Pr\{\overline{F}\}$$

$$= Pr\{E|F\}Pr\{F\} + Pr\{E|\overline{F}\}[1 - Pr\{F\}].$$

Problem. *A*: policy holder is accident prone; A_1 : policy holder will have an accident within one year. Suppose $\Pr\{A_1|A\} = 0.4$, $\Pr\{A_1|\overline{A}\} = 0.2$ and $\Pr\{A\} = 0.3$. Compute $\Pr\{A_1\}$.

Answer. $Pr\{A_1\}$

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$$Pr{A_1} = Pr{A_1|A}Pr{A}$$

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Answer.
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Answer. $\Pr\{A_1\} = \Pr\{A_1|A\}\Pr\{A\} + \Pr\{A_1|\overline{A}\}[1 - \Pr\{A\}] = 0.26$. Similarly,

$$Pr\{T|h\} = Pr\{T|F,h\}Pr\{F|h\} + Pr\{T|\overline{F},h\}Pr\{c|h\}.$$

Determine the strategy of Row at round eight if the history of play is

$$h = TC, BC, TL, BL, TR, TR, TL.$$

$$E[T] = \frac{3}{7}0 + \frac{2}{7}1 + \frac{2}{7}4 = \frac{10}{7},$$

$$E[M] = \frac{3}{7}1 + \frac{2}{7}0 + \frac{2}{7}5 = \frac{13}{7},$$

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Problem. Row is a 10% noise (plays a random action 10% of the time at random moments) fictitious player in a normal form game with payoffs

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T	0,2	1,0	4,0
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Author: Gerard Vreeswijk. Slides last modified on May 19^{th} , 2021 at 16:50

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However, Row isn't a pure fictitious player, instead it randomizes 10% of the time.

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