

Evolutionary Stochastic Games



Frank Thuijsman

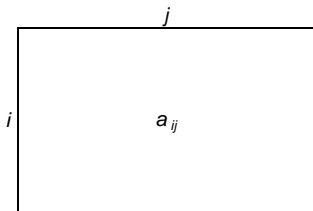
joint work with J. Flesch, T. Parthasarathy, P. Uyttendaele
Dyn Games Appl (2013) 3, 207–219

Outline

- 1 Introduction
- 2 Stochastic Games
- 3 Evolutionary Games
- 4 Evolutionary Stochastic Games
- 5 Concluding Remarks

1928, John von Neumann

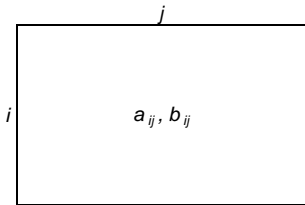
2-Person Zerosum Games



Existence of Value and Optimal Strategies

1951, John Nash

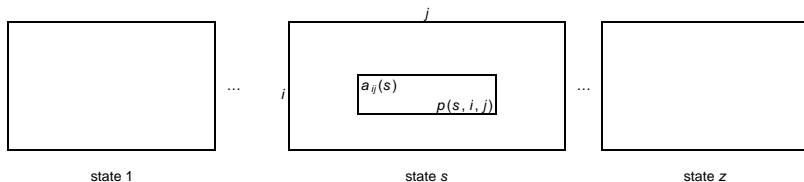
n -Person Non-Zerosum Games



Existence of Equilibria

1953, Lloyd Shapley

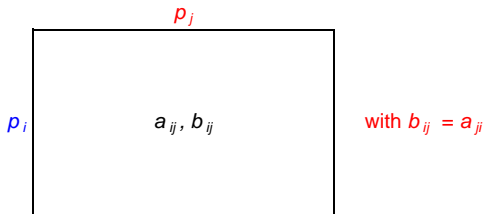
2-Person Zerosum Stochastic Games



Existence of Value and Optimal Stationary Strategies for Stopping Stochastic Games

1973, John Maynard Smith and George Price

Evolutionary Games

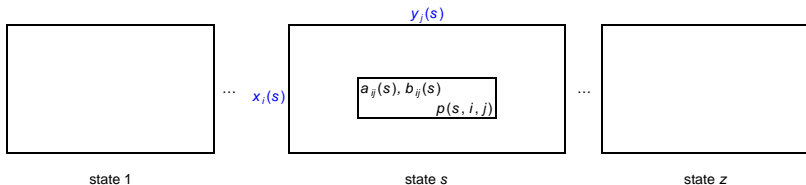


- Population consisting of Individuals of Different Types playing against Itself
- Population Distribution $p = (p_1, p_2, \dots, p_n)$
- Individuals of Type k have Fitness $e_k A p^\top$ in Population p
- Concept of Evolutionarily Stable Strategies (ESS)

Question

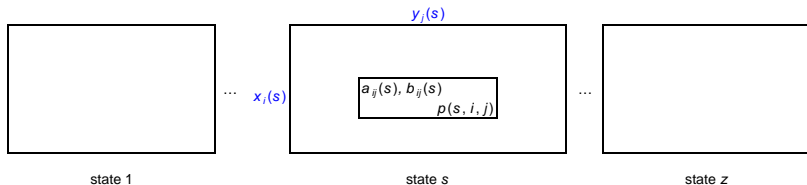
What if the Fitness of Population Members corresponds to the Average Rewards in a Stochastic Game rather than to the Expected Payoffs in a One-Shot Game?

The Stochastic Game Model



- Finitely Many States, Finitely Many Actions for each Player
- Payoffs and Transitions at each Stage 1, 2, 3, 4, ...
- Each State can serve as Initial State
- Complete Information and Perfect Recall
- Discounting or Averaging the Stage Payoffs

Some Highlights of Stochastic Game Theory



- 1953, L.S. Shapley:
2-Person Zerosum Stopping Stochastic Games - Value
- 1957, H. Everett / D. Gillette:
2-Person Zerosum Undiscounted Stochastic Games
- 1964, A.M. Fink / M. Takahashi:
 n -Person β -Discounted Stochastic Games - Equilibria
- 1981, J.-F. Mertens and A. Neyman:
2-Person Zerosum Undiscounted Stochastic Games - Value
- 2000, N. Vieille:
2-Person Undiscounted Stochastic Games - ε -Equilibria

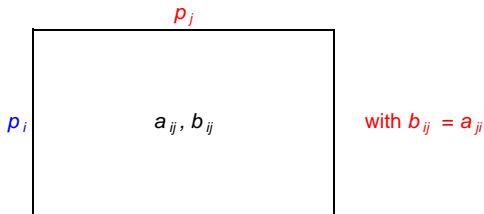
Abraham Neyman - A Personal Account



Frank Thuijsman, DKE, Maastricht University

1973, John Maynard Smith and George Price

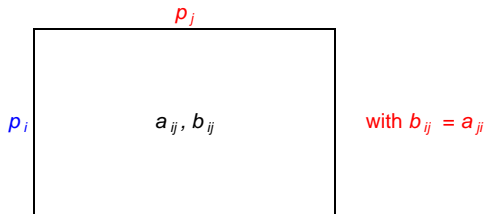
Evolutionary Games



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The ESS Concept

Evolutionary Games

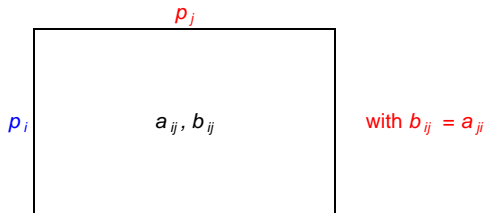


ESS: Population Distribution $p = (p_1, p_2, \dots, p_n)$ with

- $pAp^T \geq qAp^T \quad \forall q$
- If $q \neq p$ and $qAp^T = pAp^T$, then $pAq^T > qAq^T$

The Replicator Dynamic by Taylor and Jonker, 1978

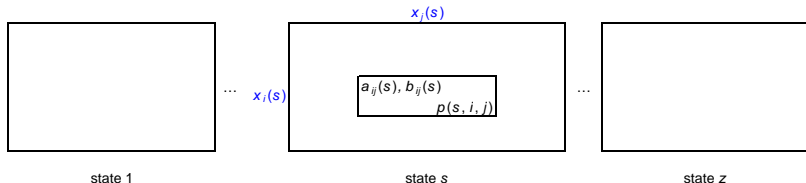
Evolutionary Games



Population Development by the Replicator Equation:

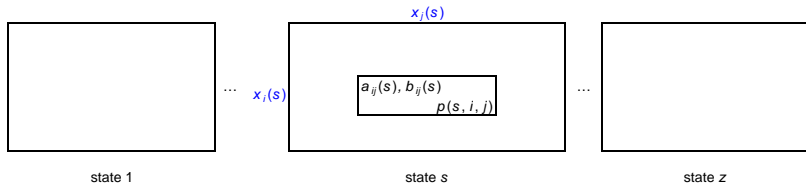
- $\dot{p}_k = p_k (e_k A p^\top - p A p^\top)$

Assumptions for Evolutionary Stochastic Games



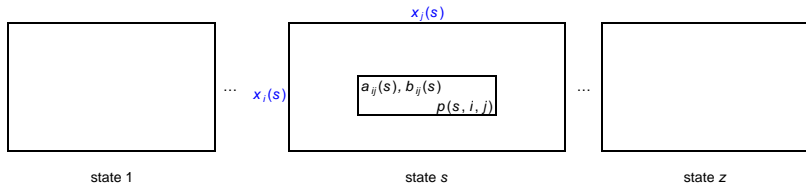
- Population playing against Itself
- *Types now correspond to Pure Stationary Strategies*
- Symmetric Payoffs: $b_{ij} = a_{ji}$
- Symmetric Transitions: $p(s, i, j) = p(s, j, i)$
- Irreducible Stochastic Game
All States communicate for all Stationary Strategies

Assumptions Continued



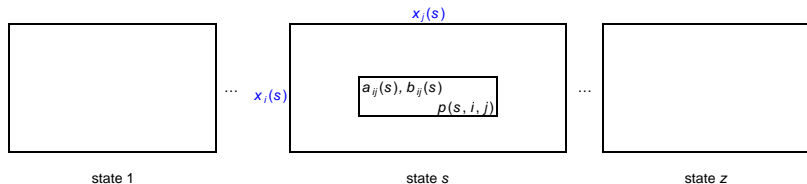
- Fitness of Individual of Type k in Population Distribution $\bar{x} = (\bar{x}_1, \bar{x}_2, \dots, \bar{x}_n)$, taken over Pure Stationary Strategies, is Average Reward $\gamma(e_k, x)$, where x is Stationary Strategy induced by \bar{x}

Assumptions Continued



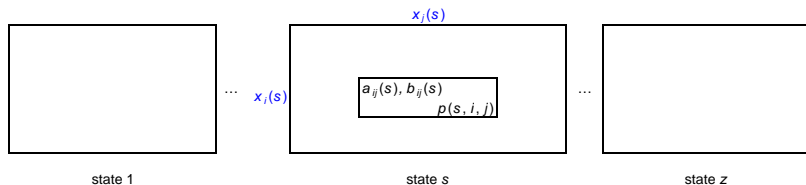
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- Different Populations can give *Same* Stationary Strategy

Assumptions Continued



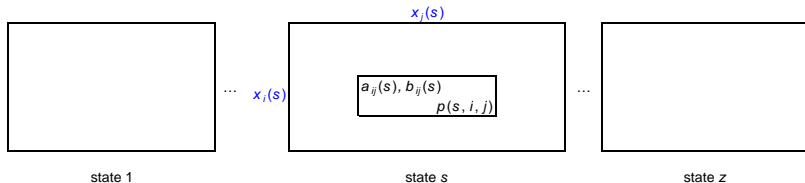
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- Different Populations can give *Same* Stationary Strategy
- *Stationary Strategy* x is ESS if
 - $\gamma(x, x) \geq \gamma(y, x) \quad \forall y$
 - If $y \neq x$ and $\gamma(y, x) = \gamma(x, x)$, then $\gamma(x, y) > \gamma(y, y)$

Assumptions Continued



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- Population Development by Replicator Dynamic
 - $\dot{\bar{x}}_k = \bar{x}_k (\gamma(e_k, x) - \gamma(x, x))$

Some Remarks



- ESS Not Always Exists
- Replicator Dynamic Not Always Converges
- Limit Points of Dynamic Not Always give ESS

Introduction
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Stochastic Games
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Evolutionary Games
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Evolutionary Stochastic Games
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Concluding Remarks
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Time Scale Assumption

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- We assume that during Individual Life Time All States are visited Sufficiently Often
- The Infinite Horizon Average Reward approximates the Finite Horizon Average Reward

Introduction
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Stochastic Games
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Evolutionary Games
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Evolutionary Stochastic Games
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Concluding Remarks
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Multi-State Interpretation

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- The Individual Fitness depends on the Actions taken at *Multiple Situations* encountered in Life *Altogether*

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- The Individual Fitness depends on the Actions taken at *Multiple Situations* encountered in Life *Altogether*
- The Fraction of Time that *Specific Situations* govern Individual Life, *depends* on the Actions taken by *All* Population Members

Different Populations, Same Stationary Strategy

In a Game with 2 States and in each State 2 Actions, T and B :

$$\frac{1}{2} \cdot (T, T) + \frac{1}{2} \cdot (B, B) = \left(\left(\frac{1}{2}, \frac{1}{2} \right), \left(\frac{1}{2}, \frac{1}{2} \right) \right) = \frac{1}{2} \cdot (T, B) + \frac{1}{2} \cdot (B, T)$$

Question

Suppose $x = \left(\left(\frac{1}{2}, \frac{1}{2} \right), \left(\frac{1}{2}, \frac{1}{2} \right) \right)$ is an ESS, does it imply that any Pure Stationary Strategy in $C(x)$ is also a Best Reply to x ?

In other words:

Do $\gamma((T, T), x)$, $\gamma((B, B), x)$, $\gamma((T, B), x)$, $\gamma((B, T), x)$
all equal $\gamma(x, x)$?

Answer

Yes

Irreducible Markov Decision Problems

Theorem

If x^ is a Stationary Optimal Strategy in an Irreducible MDP and x is a Stationary Strategy with $C(x_s) \subset C(x_s^*)$ for each s , then x is Optimal as well.*

Sketch of Proof:

We show that for a Stationary Strategy in an Irreducible MDP the Average Reward is a Convex Combination of the Average Rewards for the Pure Stationary Strategies in its Carrier. \square

Evolutionary Stochastic Games

Corollary

If x^ is an ESS in an Evolutionary Stochastic Game and $x \neq x^*$ is a Stationary Strategy with $C(x_s) \subset C(x_s^*)$ for each s , then x is no ESS.*

Proof:

If x^* ESS and $C(x_s) \subset C(x_s^*) \forall s$,

then $\gamma(x, x^*) = \gamma(x^*, x^*)$ by previous Theorem,

then $\gamma(x^*, x) > \gamma(x, x)$ by 2nd ESS Condition for x^* ,

which implies x is no ESS by 1st ESS Condition for x . \square

Another Reason for Irreducibility

Example

3, 3 (1, 0)	5, 4 (1, 0)
4, 5 (1, 0)	2, 2 (0, 1)

state 1

2, 2 (0, 1)

state 2

No Symmetric Equilibrium in Stationary Strategies
 Not even Symmetric ε -Equilibrium in Stationary Strategies

Another Reason for Irreducibility

Example

3, 3 (1, 0)	5, 4 (1, 0)	
4, 5 (1, 0)	2, 2 (0, 1)	2, 2 (0, 1)
	state 1	state 2

No Symmetric Equilibrium in Stationary Strategies
 Not even Symmetric ε -Equilibrium in Stationary Strategies
 But *there is* a Symmetric ε -Equilibrium

Theorem

For every Symmetric Irreducible Stochastic Game there exists a Symmetric Stationary Equilibrium.

For Discounted as well as for Undiscounted (Average) Rewards

Sketch of Proof:

This follows by applying a Fixed Point Argument for the Discounted Best Reply Map and by taking the Limit of Discounted Fixed Points for the Undiscounted Case. \square

A 2 State Example with Replicator Dynamics

1, 1 (0, 1)	4, 3 (.5, .5)
3, 4 (.5, .5)	2, 2 (0, 1)

state 1

3, 3 (1, 0)	5, 4 (.5, .5)
4, 5 (.5, .5)	2, 2 (1, 0)

state 2

(Trajectory)

A 2 State Example with Replicator Dynamics

1, 1 (0, 1)	4, 3 (.5, .5)
3, 4 (.5, .5)	2, 2 (0, 1)

state 1

3, 3 (1, 0)	5, 4 (.5, .5)
4, 5 (.5, .5)	2, 2 (1, 0)

state 2

(Trajectory)

A 3 State Example with Replicator Dynamics

1, 1 (.5, 0, .5)	4, 3 (.5, .5, 0)
3, 4 (.5, .5, 0)	2, 2 (0, .5, .5)

state 1

3, 3 (1, 0, 0)	5, 4 (.5, 0, .5)
4, 5 (.5, 0, .5)	2, 2 (0, 0, 1)

state 2

4, 4 (0, 1, 0)	6, 7 (0, .5, .5)
7, 6 (0, .5, .5)	5, 5 (1, 0, 0)

state 3

(Trajectory)

About this Model

Further Research:

- Finding Real Life Applications that Fit
- Exploring the Existence of Symmetric (ε -)Equilibria for Symmetric Stochastic Games

Local Interactions and Fitnesses in Space/Time

$$\begin{array}{c}
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 \end{array}
 \begin{pmatrix}
 & \blacksquare & \blacksquare & \blacksquare \\
 \blacksquare & 1 & 2 & 3 + 1.5 \sin(\pi(0.5t + y)/40) \\
 \blacksquare & 2 & 1.1 & 2 \\
 \blacksquare & 3 - 1.5 \sin(\pi(0.5t + y)/40) & 2 & 1
 \end{pmatrix}$$

GAMES 2016 & EC'16

5-th World Congress of the Game Theory Society

17-th ACM Conference on Economics and Computation: EC'16



Maastricht, 24-28 July 2016

Thanks!

