# Introduction to Evolutionary Games - 4 Escuela de Bioestocástica

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Lecture 4: Evolutionary Games

# **Replicator dynamics**

We will study general competition-collaboration dynamics with more than 2 species

- Denote our species by 1,..., m
- ▶  $x_i(t)$  denotes the abundance of species *i*, and x(t) denotes the vector  $(x_1(t), x_2(t), \ldots, x_m(t))$ .
- We consider a payoff matrix A, where A<sub>ij</sub> is the payoff that receives when i meets j
- The fitness if species i is given by

$$f_i = \sum_{j=1}^m A_{ij} x_j = (Ax)_i$$

then

$$\dot{x}_i = (f_i - \phi) x_i$$

Interpretation:

•  $f_i x_i = \sum_{j=1}^m x_i x_j A_{ij}$  is the amount of fitness that species *i* gets due to interaction with other species •  $\phi$  only help us to keep the dynamic in the simplex.

$$\phi = x^{\mathsf{T}} A x = \sum_{ij} x_i x_j A_{ij}$$

We can write the dynamic as

$$\dot{x}_i = ((Ax)_i - x^{\mathsf{T}}Ax)x_i \qquad i \in \{1, \ldots, m\}$$

► This is the so-called replicator dynamic... they always stay in the simplex

## The two-players game

A two-player game

- We have two players: the player and the house
- Each of them has to distribute 1 unit of mass over the *m* species
- The house choose a distribution y and the player a distribution x
- Reward: the Player receives

$$\mathbf{x}^{\mathsf{T}} A \mathbf{y} = \sum_{i=1}^{m} \sum_{j=1}^{m} \mathbf{x}_{i} \mathbf{y}_{j} A_{ij}$$

► Goal The player wants to maximise x<sup>T</sup>Ay

A simple strategy for the Player is just to copy the House.

– Nash Equilibrium –––––

A point y is a Nash Equilibrium if

$$x^{\mathsf{T}} A y \leq y^{\mathsf{T}} A y$$

for all  $x \in S_m$  (the simplex of *m* points).

This means that if *y* is a Nash Equilibrium, the **copying-strategy** will give the best possible reward for the player

Nash Equilibrium ——

A point y is a Nash Equilibrium if

$$x^{\mathsf{T}} A y \leq y^{\mathsf{T}} A y$$

for all  $x \in S_m$  (the simplex of *m* points).

A point y is a strict Nash equilibrium if

$$x^{\mathsf{T}} A y \leq y^{\mathsf{T}} A y$$

for all  $x \in S_m$  and  $x \neq y$ 

**Example**: Consider  $A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ . Then  $f_1 = x_1$ , and  $f_2 = x_2$ ,  $\phi = x^T A x = x_1^2 + x_2^2$ , and the whole dynamic is

dynamic is

$$\dot{x}_1 = (x_1 - \phi)x_1$$
  
 $\dot{x}_2 = (x_2 - \phi)x_2$ 

using that  $x_1 + x_2 = 1$  we have

$$\dot{x}_1 = x_1(1 - x_1)(2x_1 - 1),$$

thus the equilibrium points are (1, 0), (0, 1) and (1/2, 1/2).

– Nash Equilibrium ———

A point y is a Nash Equilibrium if

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for all  $x \in S_m$  (the simplex of *m* points).

A point y is a strict Nash equilibrium if

$$x^{\mathsf{T}} A y < y^{\mathsf{T}} A y$$

for all  $x \in S_m$  and  $x \neq y$ 

Now, let's play a game. I am the House and give you (the Player) a vector y in the simplex  $S_2$ . What do you play in order to maximise  $x^TAy = x_1y_1 + x_2y_2$ ?

- 1. If  $y_1 > y_2$  choose (1, 0)
- 2. If  $y_1 < y_2$  choose (0, 1)
- 3. if  $y_1 = y_2$  choose any vector

We can see that the Nash equilibrium are (1, 0), (0, 1) and (1/2, 1/2). **Why?** Because *y* is a Nash equilibrium if *y* is one of the **best responses**. Moreover, (1, 0) and (0, 1) are **strict** Nash Equilibrium.

## Coincidence?

#### Theorem

A Nash equilibrium y of the game described by the payoff matrix A, then y is an equilibrium point of the Replicator Dynamic associated with A.

(Technical Result\*): Moreover, if y is the  $\omega$ -limit of an orbit in the interior of the simplex  $S_n$ , then y is a Nash equilibrium.

Evolutionary Stable State \_\_\_\_\_

A point  $y \in S_m$  is an evolutionary stable state (ESS) if

 $x^{\mathsf{T}}Ax < y^{\mathsf{T}}Ax$ ,  $\forall x \neq y$  in a neighbourhood of y,

i.e. deviations from y always result in a worse payoff.

An ESS is a Nash Equilibrium, but the converse is not true.

### Theorem

If  $y \in S_m$  is an ESS, then y is an asymptotically stable rest point. Moreover, if  $y_i > 0$  for all i, then y is a globally stable rest point.

- rest point:  $y \in S_m$  is a rest point if  $f_1(y) = f_2(y) = \ldots = f_m(y)$
- stable: if we start the dynamic near y we will go to y
- globally: starting from any point x(0) with  $x_i(0) > 0$ , then the dynamics converges to y

Recall the Hawks and Doves dynamic

	Н	D
Н	$\frac{1-C}{2}$	1
D	0	1/2

For extra simplicity, fix C = 2, which we know enters in the regime that hawks are a danger for themselves. In this case

	Н	D
Н	$-\frac{1}{2}$	1
D	0	1/2

From the previous lecture, we know that  $(x_H, x_D) = (1/C, (C - 1)/C)$  is an equilibrium point of this dynamic. In our case it is just (0.5, 0.5)

Is y = (0.5, 0.5) an ESS?

	Н	D
Н	$-\frac{1}{2}$	1
D	0	1/2

We just has to verify that  $y^{T}Ax > x^{T}Ax$  for all x in a vecinity of y.

What is the vecinity of *y*? Just take  $x = (0.5 + \delta, 0.5 - \delta)$ , and let's consider all values  $|\delta|$  super small, like smaller than 0.001 (or any small value)

Then, we have

$$y^{\mathsf{T}}Ax - x^{\mathsf{T}}Ax = \delta^2,$$

which is clearly positive for any  $\delta$  (and particularly for delta close to 0) We conclude

- 1. (0.5, 0.5) is an ESS in the interior of the simplex
- 2. Then it is a globally stable rest point



# Dynamics can be complicated

Example \_

Let us consider the payoff matrix

$$A = \begin{bmatrix} 0 & 6 & -4 \\ -3 & 0 & 5 \\ -1 & 3 & 0 \end{bmatrix},$$

Then, an equilibrium point of the replicator equation is  $\dot{x}_i = x_i((A\mathbf{x})_i - \mathbf{x} \cdot A\mathbf{x})$  is

$$\mathbf{x}^* = (1/3, 1/3, 1/3).$$

but it is not an ESS



# Rock-Scissors-Paper and evolutionary games



# Wasn't it rock-paper-scissors?

When we establish the order Rock-Scissors-Paper, there is a linearity in terms of the winner per combination:





igure: Cachipún punctuation and payoff matrix

## Definition (Zero-sum games)

If the gain of one player is always the loss of the other, i.e. the payoff matrix A is anti-symmetric  $(A^{T} = -A)$ , then the game is called a **zero-sum game**.

For  $\epsilon = 0$ , i.e. no reward in case of a tie, the Cachipún game is a zero-sum game, for which we have

$$\mathbf{x} \cdot A\mathbf{x} = 0.$$

Notice that the RSP-payoff matrix is given by:

$$A = \begin{pmatrix} 0 & 1 & -1 \\ -1 & 0 & 1 \\ 1 & -1 & 0 \end{pmatrix}$$

What are the Nash Equilibria?

Only (1/3, 1/3, 1/3)

Exercise: Write the replicator dynamics associated with this dynamic

$$A = \begin{pmatrix} 0 & 1 & -1 \\ -1 & 0 & 1 \\ 1 & -1 & 0 \end{pmatrix}$$

Solution:

$$\begin{cases} \dot{x}_1 = x_1(x_2 - x_3) \\ \dot{x}_2 = x_2(x_3 - x_1) \\ \dot{x}_3 = x_3(x_1 - x_2) \end{cases}$$

with  $(x_1, x_2, x_3) \in S_3$ .

Solving the system  $\dot{\mathbf{x}}(t) = \mathbf{0}$  gives us the equilibrium points:

$$p_1 = (1, 0, 0)$$

$$p_2 = (0, 1, 0)$$

$$p_3 = (0, 0, 1)$$

$$\hat{\mathbf{x}} = (1/3, 1/3, 1/3).$$

Is some of these points an ESS? The answer is no.



### There is no ESS.







Rock-scissors-paper is not just a naïve example. This behaviour can be observed also in natural systems, for example, with microbial communities containing toxin-producing (or colicinogenic) E. coli. These bacteria can encode the toxin but just a small fraction o them will release the colicin.

In the above example, the cells can be divided in three **types**: resistant cells (R), colicinogenic cells (C) and sensitive cells (S).

- The growth rate of **R** cells will exceed that of **C** cells since they avoid the competitive cost of carrying the col plasmid,
- R cells suffer because colicin is also involved in crucial cell functions such as nutrient uptake, so they growth rate will be less than the growth rate of S cells.
- colicin-sensitive bacteria are killed by the colicin, although may occasionally experience mutations that render them resistant to the colicin.



### Who would you choose as rock, paper or scissors?

# Final words about the course

- This is just a very basic introduction. There is much more to learn
- Some books
  - 1. Novak Evolutionary Dynamics. Very introductory, but a bit too shallow and informal
  - 2. Hofbauer and Sigmund Evolutionary Games and Population Dynamics. Much harder, very formal and proof-based
- There are many resources online, but they are mostly based on the previous books (e.g. same examples etc..)
- Numerics: most software can solve differential equation: e.g. Mathematica, Matlab, probably some library in R, and many libraries in Python
- Stochastic approach: The fact that the system is stable only means that the abundances are stable. In reality mass is moving quite a lot in a sort of ballanced way. If we analyse the movement of one 'particle' it would be a random processes jumping between species.
- Some prerequisites for self-studying? More of less the same to study clasical mechanics. I reckon
  - 1. Calculus at Engeneering level is probably enough
  - 2. Linear Algebra
  - 3. Some knowledge of probability
- What about the actual values of the rates, payoff matrices, etc? We need data, and there are a lot of statistical problems here. There is a big issue: you can quickly scalate and have a lot of parameters and not a lot of data (used to be a problem in the past, but we had made progress)
- All my material is (and I will update a few things) in https://nnrivera.github.io/teaching/biostochastics2024/