



# An Introduction to Evolutionary Game Theory: Lecture 2

Mauro Mobilia

Lectures delivered at the Graduate School on Nonlinear and Stochastic Systems in Biology  
held in the Department of Applied Mathematics, School of Mathematics  
University of Leeds, U.K.

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The goal of this lecture is to give some insight into the following topics:

- Some Properties of the Replicator Dynamics
- Replicator Equations for  $2 \times 2$  Games
- Moran Process & Evolutionary Dynamics
- The Concept of Fixation Probability
- Evolutionary Game Theory in Finite Population
- Influence of Fluctuations on Evolutionary Dynamics

# Replicator Dynamics

Population of  $Q$  different species:  $\mathbf{e}_1, \dots, \mathbf{e}_Q$ , with frequencies  $x_1, \dots, x_Q$

State of the system described by  $\mathbf{x} = (x_1, \dots, x_Q) \in \mathcal{S}_Q$ , where  
 $\mathcal{S}_Q = \{\mathbf{x}; x_i \geq 0, \sum_{i=1}^Q x_i = 1\}$

To set up the dynamics, we need a functional expression for the fitness  $f_i(\mathbf{x})$

Between various possibilities, a very popular choice is:

$$\dot{x}_i = x_i (f_i(\mathbf{x}) - \bar{f}(\mathbf{x})),$$

where, one (out of many) possible choices, for the fitness is the *expected payoff*:  $f_i(\mathbf{x}) = \sum_{j=1}^Q A_{ij} x_j$

and  $\bar{f}(\mathbf{x})$  is the *average fitness*:  $\bar{f}(\mathbf{x}) = \sum_i^Q x_i f_i(\mathbf{x})$

This choice corresponds to the so-called **replicator dynamics** on which most of evolutionary game theory is centered

# Some Properties of the Replicator Dynamics (I)

Replicator equations (REs):

$$\dot{x}_i = x_i [(\mathbf{Ax})_i - \mathbf{x} \cdot \mathbf{Ax}]$$

Set of coupled cubic equations (when  $\mathbf{x} \cdot \mathbf{Ax} \neq 0$ )

Let  $\mathbf{x}^* = (x_1^*, \dots, x_Q^*)$  be a fixed point (steady state) of the REs

- $\mathbf{x}^*$  can be (Lyapunov-) stable, unstable, attractive (i.e. there is basin of attraction), asymptotically stable=attractor (=stable + attractive), globally stable (basin of attraction is  $\mathcal{S}_Q$ )
- Only possible interior fixed point satisfies (there is either 1 or 0):

$$(\mathbf{Ax}^*)_1 = (\mathbf{Ax}^*)_2 = \dots = (\mathbf{Ax}^*)_Q = \mathbf{x}^* \cdot \mathbf{Ax}^*$$

$$x_1 + \dots + x_Q = 1$$

- Same dynamics if one adds a constant  $c_j$  to the payoff matrix  
 $\mathbf{A} = (A_{ij})$ :  $\dot{x}_i = x_i [(\mathbf{Ax})_i - \mathbf{x} \cdot \mathbf{Ax}] = x_i [(\tilde{\mathbf{A}}\mathbf{x})_i - \mathbf{x} \cdot \tilde{\mathbf{A}}\mathbf{x}]$ , where  
 $\tilde{\mathbf{A}} = (A_{ij} + c_j)$

# Some Properties of the Replicator Dynamics (II)

Dynamic versus evolutionary stability: connection between dynamic stability (of REs) and NE/evolutionary stability?

*Notions do not perfectly overlap  $\Rightarrow$  Folks Theorem of EGT:*

Let  $\mathbf{x}^* = (x_1^*, \dots, x_Q^*)$  be a fixed point (steady state) of the REs

- NEs are rest points (of the REs)
- Strict NEs are attractors
- A stable rest point (of the REs) is an NE
- Interior orbit converges to  $\mathbf{x}^* \Rightarrow \mathbf{x}^*$  is an NE
- ESSs are attractors (asymptotically stable)
- Interior ESSs are global attractors

**Converse statements generally do not hold!**

- For  $2 \times 2$  matrix games  $\mathbf{x}^*$  is an ESS iff it is an attractor
- REs with  $Q$  strategies can be mapped onto Lotka-Volterra equations for  $Q - 1$  species:  $\dot{y}_i = y_i \left( r_i + \sum_{j=1}^{Q-1} b_{ij} y_j \right)$
- Replicator dynamics is non-innovative: cannot generate new strategies

# Replicator Dynamics for $2 \times 2$ Games (I)

2 strategies: say  $A$  and  $B$

$N$  players:  $N_A$  are  $A$ -players and  $N_B$  are  $B$ -players,  $N_A + N_B = N$

General payoff matrix:

vs	$A$	$B$
$A$	$1 + p_{11}$	$1 + p_{12}$
$B$	$1 + p_{21}$	$1 + p_{22}$

where selection  $\rightarrow p_{ij}$  and the neutral component  $\rightarrow 1$

Frequency of  $A$  and  $B$  strategists is resp.

$$x = N_A/N \quad \text{and} \quad y = N_B/N = 1 - x$$

Fitness (expected payoff) of  $A$  and  $B$  strategists is resp.

$$f_A(x) = p_{11}x + p_{12}(1 - x) + 1 \quad \text{and} \quad f_B(x) = p_{21}x + p_{22}(1 - x) + 1$$

Average fitness:  $\bar{f}(x) = xf_A(x) + (1 - x)f_B(x)$

# Replicator Dynamics for $2 \times 2$ Games (II)

Replicator dynamics:

$$\begin{aligned}\frac{dx}{dt} &= x[f_A(x) - \bar{f}(x)] = x(1-x)[f_A(x) - f_B(x)] \\ &= x(1-x)[x(p_{11} - p_{21}) + (1-x)(p_{12} - p_{22})]\end{aligned}$$

- $xy = x(1-x)$ : interpreted as the probability that **A** and **B** interact
- $f_A(x) - f_B(x) = x(p_{11} - p_{12}) + (1-x)(p_{12} - p_{22})$ : says that reproduction (“success”) depends on the difference of fitness

Equivalent payoff matrix ( $A_{i1} \rightarrow A_{i1} - p_{11}$ ,  $A_{i2} \rightarrow A_{i2} - p_{22}$ ), with  $\mu_A = p_{21} - p_{11}$  and  $\mu_B = p_{12} - p_{22}$ :

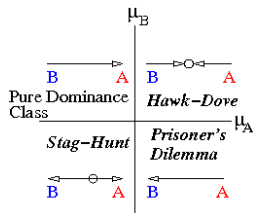
vs	<b>A</b>	<b>B</b>
<b>A</b>	1	$1 + \mu_A$
<b>B</b>	$1 + \mu_B$	1

$$\frac{dx}{dt} = x(1-x)[-x\mu_A + (1-x)\mu_B] = x(1-x)[\mu_B - (\mu_A + \mu_B)x]$$

⇒ For  $2 \times 2$  games, the dynamics is simple: no limit cycles, no oscillations, no chaotic behaviour

# Replicator Dynamics for $2 \times 2$ Games (III)

$$\frac{dx}{dt} = x(1-x)[\mu_B - (\mu_A + \mu_B)x]$$



- $\mu_A > 0$  and  $\mu_B > 0$ : Hawk-Dove game  
 $x^* = \frac{\mu_B}{\mu_A + \mu_B}$  is stable (attractor, ESS) interior FP
- $\mu_A > 0$  and  $\mu_B < 0$ : Prisoner's Dilemma  
 $B$  always better off,  $x^* = 0$  is ESS
- $\mu_A < 0$  and  $\mu_B < 0$ : Stag-Hunt Game  
 Either  $A$  or  $B$  can be better off, i.e.  $x^* = 0$  and  $x^* = 1$  are ESS.  
 $x^* = \frac{\mu_B}{\mu_A + \mu_B}$  is unstable FP (non-ESS)
- $\mu_A < 0$  and  $\mu_B > 0$ : Pure Dominance Class  
 $A$  always better off,  $x^* = 1$  is ESS



# Some Remarks on Replicator Dynamics

$$\frac{dx}{dt} = x(1-x)[\mu_B - (\mu_A + \mu_B)x]$$

- For  $x$  small:  $\dot{x} = \mu_B x$
- For  $x \approx 1$ :  $\dot{y} = (d/dt)(1-x) = \mu_A(1-x)$

Thus, the stability of  $x^* = 0$  and  $x^* = 1$  simply depends on the sign of  $\mu_B$  and  $\mu_A$ , respectively

Another popular dynamics is the so-called “adjusted replicator dynamics”, for which the equations read:

$$\frac{dx}{dt} = x \frac{f_A(x) - \bar{f}(x)}{\bar{f}(x)} = x(1-x) \left[ \frac{f_A(x) - f_B(x)}{\bar{f}(x)} \right]$$

These equations share the same fixed points with the REs. In general, replicator dynamics and adjusted replicator dynamics give rise to different behaviours. However, for  $2 \times 2$  games: same qualitative behaviour

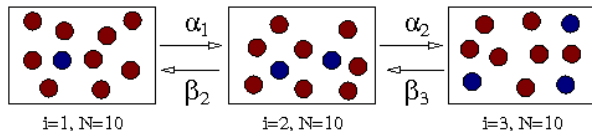
# Stochastic Dynamics & Moran Process

Evolutionary dynamics involves a *finite number of discrete individuals*  
⇒ “Microscopic” stochastic rules given by the **Moran process**

Moran Process is a Markov birth-death process in 4 steps:

2 species,  $i$  individuals of species  $A$  and  $N - i$  of species  $B$

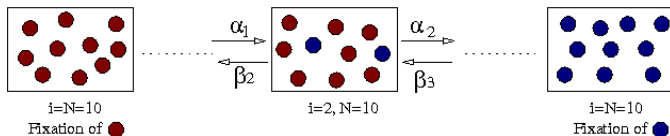
- 1 An individual  $A$  could be chosen for *birth and death* with probability  $(i/N)^2$ . The number of  $A$  remains the same
- 2 An individual  $B$  could be chosen for *birth and death* with probability  $((N - i)/N)^2$ . The number of  $B$  remains the same
- 3 An individual  $A$  could be chosen for *reproduction* and a  $B$  individual for *death* with probability  $i(N - i)/N^2$ . For this event:  
 $i \rightarrow i + 1$  and  $N - i \rightarrow N - 1 - i$
- 4 An individual  $B$  could be chosen for *reproduction* and a  $A$  individual for *death* with probability  $i(N - i)/N^2$ . For this event:  
 $i \rightarrow i - 1$  and  $N - i \rightarrow N + 1 - i$



# Stochastic Dynamics & Moran Process

Evolutionary dynamics given by the **Moran process**: Markov birth-death process in 4 steps

There are two absorbing states in the Moran process: **all-B** and **all-A**



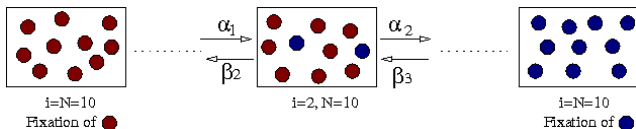
What is the probability  $F_i$  of ending in a state with all **A** ( $i = N$ ) starting from  $i$  individuals **A**? For  $i = 1$ ,  $F_1$  is the “fixation” probability of **A**

- Transition from  $i \rightarrow i + 1$  given by rate  $\alpha_i$
- Transition  $i \rightarrow i - 1$  given by rate  $\beta_i$

$$F_i = \beta_i F_{i-1} + (1 - \alpha_i - \beta_i) F_i + \alpha_i F_{i+1}, \quad \text{for } i = 1, \dots, N - 1$$

$$F_0 = 0 \quad \text{and} \quad F_N = 1$$

# Moran Process & Fixation Probability



What is the fixation probability  $F_1$  of  $A$  individuals?

$$F_i = \beta_i F_{i-1} + (1 - \alpha_i - \beta_i) F_i + \alpha_i F_{i+1}, \quad \text{for } i = 1, \dots, N-1$$

$$F_0 = 0 \quad \text{and} \quad F_N = 1$$

Introducing  $g_i = F_i - F_{i-1}$  ( $i = 1, \dots, N-1$ ), one notes that  $\sum_{i=1}^N g_i = 1$  and  $g_{i+1} = \gamma_i g_i$ , where  $\gamma_i = \beta_i / \alpha_i \Rightarrow$  one recovers a classic results on

Markov chains: 
$$F_i = \frac{1 + \sum_{j=1}^{i-1} \prod_{k=1}^j \gamma_k}{1 + \sum_{j=1}^{N-1} \prod_{k=1}^j \gamma_k}$$

$\Rightarrow$  Fixation probability of species  $A$  is  $F_A = F_1 = \frac{1}{1 + \sum_{j=1}^{N-1} \prod_{k=1}^j \gamma_k}$

As  $i = 0$  and  $i = N$  are absorbing states  $\Rightarrow$

always *absorption* (all- $A$  or all- $B$ )  $\Rightarrow$  Fixation probability of species  $B$

is  $F_B = 1 - F_{N-1} = \frac{\prod_{k=1}^{N-1} \gamma_k}{1 + \sum_{j=1}^{N-1} \prod_{k=1}^j \gamma_k}$

# Fixation in the Neutral & Constant Fitness Cases

## Fixation Probabilities:

$$F_A = \frac{1}{1 + \sum_{j=1}^{N-1} \prod_{k=1}^j \gamma_k} \text{ and } F_B = F_A \prod_{k=1}^{N-1} \gamma_k, \text{ with } \gamma_i = \beta_i / \alpha_i$$

- When  $\alpha_i = \beta_i = \gamma_i = 1$ , this is the neutral case where there is no *selection* but only *random drift*:

$$F_A = F_B = 1/N$$

This means that the chance that an individual will generate a lineage which will inheritate the entire population is  $1/N$

- Case where  $A$  and  $B$  have constant but different fitnesses,  $f^A = r$  for  $A$  and  $f^B = 1$  for  $B$ ,  $\alpha_i = \frac{ri(N-i)}{N(N+(r-1)i)}$  and  $\beta_i = \frac{i(N-i)}{N(N+(r-1)i)}$

$$\text{Thus, } F_A = \frac{1-r^{-1}}{1-r^{-1}N} \text{ and } F_B = \frac{1-r}{1-rN}$$

If  $r > 1$ ,  $F_A > N^{-1}$  for  $N \gg 1$ : selection favours the fixation of  $A$

If  $r < 1$ ,  $F_B > N^{-1}$  for  $N \gg 1$ : selection favours the fixation of  $B$

# Evolutionary Games in Finite Populations (I)

Finite population of 2 species:  $i$  individuals of species  $A$  and  $N - i$  individuals of species  $B$  interact according to the payoff matrix:

vs	$A$	$B$
$A$	$a$	$b$
$B$	$c$	$d$

Probability to draw a  $A$  and  $B$  is  $i/N$  and  $(N - i)/N$ , respectively  $\Rightarrow$

- Probability that a given individual  $A$  interacts with another  $A$  is  $(i - 1)/(N - 1)$
- Probability that a given individual  $A$  interacts with a  $B$  is  $(N - i)/(N - 1)$
- Probability that a given individual  $B$  interacts with another  $B$  is  $(N - i - 1)/(N - 1)$
- Probability that a given individual  $B$  interacts with a  $A$  is  $i/(N - 1)$

The states  $i = 0$  (All- $A$ ) and  $i = N$  (All- $B$ ) are absorbing

Expected payoff for  $A$  and  $B$ , respectively:

$$E_i^A = \frac{a(i - 1) + b(N - i)}{N - 1} \quad \text{and} \quad E_i^B = \frac{ci + d(N - i - 1)}{N - 1}$$

# Evolutionary Games in Finite Populations (II)

$$F_A = \left(1 + \sum_{j=1}^{N-1} \prod_{k=1}^j \gamma_k\right)^{-1} \quad \text{and} \quad F_B = F_A \prod_{k=1}^{N-1} \gamma_k, \quad \text{with } \gamma_i = \beta_i / \alpha_i$$
$$E_i^A = \frac{a(i-1) + b(N-i)}{N-1} \quad \text{and} \quad E_i^B = \frac{ci + d(N-i-1)}{N-1}$$

Expected payoffs  $E_i^{A,B}$  are usually interpreted as fitness.

Recent idea (Nowak et al.): Introduce a parameter  $w$  accounting for background random drift contribution to fitness  $f_i^A$  for  $A$  and  $f_i^B$  for  $B$

$$f_i^A = 1 - w + wE_i^A \quad \text{and} \quad f_i^B = 1 - w + wE_i^B$$

Average fitness:  $\bar{f} = (i/N)f_i^A + (1 - (i/N))f_i^B$

Parameter  $w$  measures the *intensity of selection*:  $w = 0 \Rightarrow$  no selection (only random drift),  $w = 1 \Rightarrow$  only selection,  $w \ll 1 \Rightarrow$  “weak selection”

Consider a Moran process with frequency-dependent hopping rates:

$$\alpha_i = \frac{f_i^A}{\bar{f}} \binom{i}{N} \binom{N-i}{N} \quad \text{and} \quad \beta_i = \frac{f_i^B}{\bar{f}} \binom{i}{N} \binom{N-i}{N} \Rightarrow \gamma_i = \frac{f_i^B}{f_i^A}$$

Thus,  $F_A = 1 / \left(1 + \sum_{j=1}^{N-1} \prod_{k=1}^j (f_k^B / f_k^A)\right)$  and  $F_B = F_A \prod_{k=1}^{N-1} (f_k^B / f_k^A)$

## Fixation Probabilities:

$$F_A = \left( 1 + \sum_{j=1}^{N-1} \prod_{i=1}^j \frac{f_i^B}{f_i^A} \right)^{-1} \quad \text{and} \quad F_B = F_A \prod_{i=1}^{N-1} (f_i^B / f_i^A), \quad \text{with}$$

$$f_i^A = 1 - w + w \frac{a(i-1) + b(N-i)}{N-1} \quad \text{and} \quad f_i^B = 1 - w + w \frac{ci + d(N-i-1)}{N-1}$$

**Does selection favour fixation of  $A$ ? Yes, only if  $F_A > 1/N$**

In the weak selection limit ( $w \rightarrow 0$ ):

$$F_A \approx \frac{1}{N} \left[ 1 - \frac{w}{6} (\{a + 2b - c - 2d\} N - \{2a + b + c - 4d\}) \right]^{-1}$$

Thus,  $F_A > 1/N$  if  $a(N-2) + b(2N-1) > c(N+1) + 2d(N-2)$

$$\begin{array}{l|l} N=2 & b > c \\ N=3 & a+5b > 2(2c+d) \\ N=4 & 2a+7b > 5c+4d \\ \dots & \dots \\ N \gg 1 & a+2b > c+2d \end{array}$$

For large  $N$ ,  $F_A > 1/N$  if  $a+2b > c+2d$



In the weak selection limit ( $w \rightarrow 0$ ):

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For large  $N$ ,  $F_A > 1/N$  if  $a+2b > c+2d$

## Consequences of finite-size fluctuations?

Reconsider a  $2 \times 2$  game with  $a > c$  and  $b < d$  (“Stag-Hunt game”):

- Rational game: **all-A** and **all-B** are strict-NE and ESS
- Replicator Dynamics: **all-A** & **all-B** attractors and  $x^* = \frac{d-b}{a-c+d-b}$  is an unstable interior rest point (NE, but not ESS)
- In finite (yet large) population (stochastic Moran process, weak selection): The condition  $a - c > 2(d - b)$  to favour fixation of **A** leads to  $x^* < 1/3$

- If the unstable rest point  $x^*$  occurs at frequency  $< 1/3$ , in a large yet finite population and for  $w \ll 1$ , selection favours the fixation of **A**  
- Probability that a single **A** takes over the entire population of  $N - 1$  individuals **B** is greater than  $1/N$

# Influence of Fluctuations & Finite-Size Effects (II)

In the weak selection limit ( $w \rightarrow 0$ ):

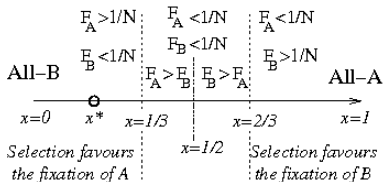
$$F_A \approx \frac{1}{N} \left[ 1 - \frac{w}{6} (\{a+2b-c-2d\}N - \{2a+b+c-4d\}) \right]^{-1}$$

For large  $N$ ,  $F_A > 1/N$  if  $a+2b > c+2d$

## Consequences of finite-size fluctuations?

Reconsider a  $2 \times 2$  game with  $a > c$  and  $b < d$  ("Stag-Hunt game"):

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- In finite (yet large) population (stochastic Moran process, weak selection): The condition  $a - c > 2(d - b)$  to favour fixation of **A** leads to  $x^* < 1/3$



# Influence of Fluctuations & Finite-Size Effects (III)

In the weak selection limit ( $w \rightarrow 0$ , 2 species systems):

$$F_A \approx \frac{1}{N} \left[ 1 - \frac{w}{6} (\{a+2b-c-2d\}N - \{2a+b+c-4d\}) \right]^{-1}$$

Previous result hints that the concept of *evolutionary stability* should be modified to account for finite-size fluctuations  $\Rightarrow$  leads to the concept of **ESS<sub>N</sub>**: A finite population of **B** is evolutionary stable is evolutionary stable against a second species **A** if

- 1 The fitness of **B** is greater than that of **A**, i.e.  $f_i^B > f_i^A, \forall i$ . This means: “selection opposes **A** invading **B**”
- 2  $F_A < 1/N$ , implying that selection opposes **A** replacing **B**

This leads to the **criteria for evolutionary stability** of **B**:

Deterministic ( $N = \infty$ )	Stochastic ( $N$ finite)
(1) $d > b$	$(d - b)N > 2d - (b + c)$
(2) if $b = d$ , then $c > a$	$c(N + 1) + 2d(N - 2) > a(N - 2) + b(2N - 1)$

# Influence of Fluctuations & Finite-Size Effects (III)

In the weak selection limit ( $w \rightarrow 0$ ,  $2 \times 2$  games):

$$F_A \approx \frac{1}{N} \left[ 1 - \frac{w}{6} (\{a+2b-c-2d\}N - \{2a+b+c-4d\}) \right]^{-1}$$

Criteria for evolutionary stability of  $B$  in a population of size  $N$ :

Deterministic ( $N = \infty$ )	Stochastic ( $N$ finite)
(1) $d > b$	$(d-b)N > 2d - (b+c)$
(2) if $b = d$ , then $c > a$	$c(N+1) + 2d(N-2) > a(N-2) + b(2N-1)$

Conditions for evolutionary stability depend on the population size:

$B$ is ESS $_N$ if	$N = 2$	$N \gg 1$ (finite)
Condition (1):	$c > b$	$d > b$
Condition (2):	$c > b$	$x^* = \frac{d-b}{a-c+d-b} > 1/3$

- For small  $N$ , the traditional ESS conditions are *neither necessary nor sufficient* to guarantee evolutionary stability
- For large  $N$ , the traditional ESS conditions are *necessary but not sufficient* to guarantee evolutionary stability

In this set of lectures dedicated to an introduction to evolutionary game theory, we have discussed

- Some concepts of *classic (rational) game theory* which were illustrated by a series of examples (hawk-doves, prisoner's dilemma and stag-hunt games)
- Notion of evolutionary dynamics via the concept of fitness
- *Replicator dynamics* and discussed its properties: connection between dynamic stability, NEs and ESS
- *Replicator dynamics* for  $2 \times 2$  games: classification
- Stochastic evolutionary dynamics according to the Moran process
- Fixation probability as a Markov chain problem
- Fixation probability for (a) the neutral case, (b) the case with constant fitness, (c)  $2 \times 2$  **games with finite populations**
- *Influence of fluctuations*: fixation and new criteria for evolutionary stability ( $ESS_N$  for  $2 \times 2$  games)

Further topics and some open problems (non-exhaustive list):

- **Replicator dynamics for  $Q \times Q$  games:**
  - For  $Q \geq 3$ : Replicator equations  $\Rightarrow$  cycles, oscillations, chaos ( $Q > 3$ ), ...
  - Spatial degrees of freedom and role of mobility (PDE): pattern formation
  - $Q \times Q$  games with mutations
- **Stochastic evolutionary dynamics:**
  - Stochastic evolutionary game theory on lattices and graphs
  - Combined effects of mobility, fluctuations, and selection

For  $Q \times Q$  games, with  $Q \geq 3$ :

  - Diffusion approximation (Fokker-Planck equation)
  - Fixation and extinction times (e.g. as first-passage problems)
  - Generalization of the concept of  $ESS_N$