## Introduction to Evolutionary Games - 2 Escuela de Bioestocástica

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Here we have

$$\begin{array}{l} \dot{x} = ax \\ \dot{y} = by \end{array} \qquad \Longrightarrow \qquad x(t) = x_0 e^{at} \\ y(t) = y_0 e^{bt} \end{array}$$

Therefore,

$$\frac{x(t)}{y(t)} = \frac{x_0}{y_0} e^{(a-b)t}$$

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Therefore,

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- if a = b then the quotient is constant: the size of the population are in the same proportion
- ▶ if a > b then  $\lim_{t\to\infty} \frac{x(t)}{y(t)} \to \infty$ : Population A outcompete population B
- if a < b then  $\lim_{t\to\infty} \frac{x(t)}{v(t)} \to 0$ : Population *B* outcompete population *A*

A much more interesting case is when the total abundance is constant, i.e. we set x(t) + y(t) = 1. In this setting the simplest model is

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 $(a - \phi(t))x(t) + (b - \phi(t))y(t) = 0 \rightarrow ax + by = \phi(x + y) \rightarrow \phi = (ax + by)$ 

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Exercise

If  $x(0) \in (0, 1)$ , show that  $\lim_{t\to\infty} x(t) \to 1$  if a > b, and  $\lim_{t\to\infty} x(t) \to 0$  if b < a

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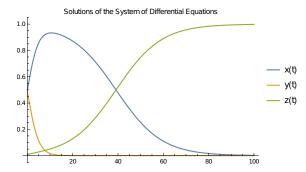
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But.... Computers!

Mathematica is my to-go software for solving stuff. Other alternatives like Mathlab or several libraries in python (scypy, numpy, etc)

```
a = 1.9; b = 1.5; c = 2; (*fitness value*)
phi[t_] := a*x[t] + b*y[t] + c*z[t]; (*average fitness*)
eq1 = (x'[t] == (a - phi[t])*x[t]); (*the three equations and initial conditions*)
eq2 = (y'[t] == (b - phi[t])*y[t]);
eq3 = (z'[t] == (c - phi[t])*z[t]);
initialConditions = {x[0] == 50/100, y[0] == 49/100, z[0] == 1/100};
solution = NDSolve[{eq1, eq2, eq3, initialConditions},
{x[t], y[t], z[t]}, {t, 0, 100}]; (*solver for ODE*)
{xSol, ySol, zSol} = {x[t], y[t], z[t]} /.
solution[[1]]; (*extract the solutions*)
Plot[{xSol, ySol, zSol}, {t, 0, 100}, PlotLegends-> {"x(t)", "y(t)", "z(t)"},
FrameLabel -> {"t", "Values"}, PlotLabel-> "Solutions of the System of Differential
```

Example with a = 1.9, b = 1.5, c = 2, x(0) = 50/100, y(0) = 49/100, z(0) = 1/100



DETOUR: Simplex and Equilibrium Points

In the previous examples, we had that the total abundance was constant.

The collection of points  $(x_1, \ldots, x_n)$  such that  $x_i \ge 0$  and  $\sum_{i=1}^n x_i = 1$  are called the simplex  $S_n$ .

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- Each point of the simplex represents the abundance of the population.
- If we keep the total abundance fixed at 1, then any evolutionary dynamics will be a dynamic on the simplex.

# A bit on equilibrium Points

When we have a system of diferential equations

$$\dot{x} = f(x, y, z)$$
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Solutions to such a system are important because they are points of "0" velocity. Those points are called **equilibria points** 

No all equilibria are the same

- 1. Stable Equilibrium: the system tends to return to that point after a perturbation
- 2. Unstable Equilibrium: the system moves away after a perturbation
- 3. Saddle Point: the system returns and moves away in different directions after a perturbation
- 4. Center: the system moves around the equilibrium
- 5. Others

## END OF DETOUR

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Let's consider again two species A and B.

- Let x and y be the abundances of A and B respectively, and assume the dynamic is on the simplex
- Suppose that A has fitness a, and mutates into B at rate  $m_{AB}$ . Similarly
- Suppose that *B* has fitness *b* and mutates into *A* at rate  $m_{BA}$ .
- We will assume that both  $m_{AB} > 0$  and  $m_{BA} > 0$

Then, the system of equations is given by

$$\dot{x} = ax - m_{AB}x + m_{BA}y - \phi x$$
$$\dot{y} = by - m_{BA}y + m_{AB}x - \phi y$$

since x + y = 1 and  $\dot{x} + \dot{y} = 0$ , we have  $\phi(t) = ax + by$ .

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**Equilibrium Point?** We shall solve  $\dot{x} = 0$ ,  $\dot{y} = 0$ . Let's solve for *x*:

$$0 = \dot{x} = ax - m_{AB}x + m_{BA}y - (ax + by)x$$

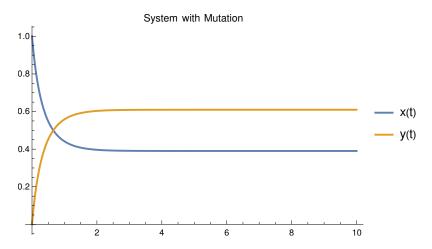
replacing y = 1 - x yields

$$(b-a)x^{2} + (a-b-(m_{AB}+m_{BA}))x + m_{BA}$$

- if a = b, then  $(m_{AB} + m_{BA})(x + m_{BA})$
- If a ≠ b, then we have a quadratic equation. The analysis of equilibrium points is harder here.

$$x
ightarrow rac{b-a+m_{AB}+m_{BA}\pm\sqrt{(b-a+m_{AB}+m_{BA})^2-4(b-a)m_{BA}}}{2(b-a)}$$

A system with a = 5, b = 3,  $m_A B = 2$ ,  $m_{BA} = 0.5$ 



# **Mutation Matrix**

Let's have a look again at our system of differential equations:

$$\dot{x} = ax - m_{AB}x + m_{BA}y - \phi x$$
$$\dot{y} = by - m_{BA}y + m_{AB}x - \phi y$$

Now consider another species, C, then we would have a system like

$$\dot{x} = ax - (m_{AB} + m_{AC})x + (m_{BA}y + m_{CA}z) - \phi x$$
  
$$\dot{y} = by - (m_{BA} + m_{BC})y + (m_{AB}x + m_{CA}z) - \phi y$$
  
$$\dot{z} = cz - (m_{CA} + m_{CB})z + (m_{AC}x + m_{BC}y) - \phi z$$

Then, we can write

$$Q = \begin{pmatrix} -(m_{AB} + m_{AC}) & m_{AB} & m_{AC} \\ m_{BA} & -(m_{BA} + m_{BC}) & m_{BC} \\ m_{CA} & m_{CB} & -(m_{CA} + m_{CB}) \end{pmatrix}$$
  
then, by using the notation  $\overrightarrow{X} = (x, y, z)$ , the mutation part can be written as

$$Q^{\intercal} \overrightarrow{X}$$

In general, any mutation matrix Q is such that

- 1. For diagonal entries:  $Q_{ii} \leq 0$
- 2. For odd-diagonal entries  $Q_{ij} \ge 0$
- 3. The sum of the entries of each column is 0.

# Summary

#### **Reproduction:**

- Reproduction can be model as ax with a > 0 being the reproduction rate, meaning that each individual reproduces at rate a
- ▶ Dead can be model as -dx meaning that each individual dies at rate d
- ▶ the simplest model is then  $\dot{x} = (a d)x$ . We can allow  $a \in \mathbb{R}$  and encode reproduction and dead in the variable *a*

#### Selection and Competition:

- We allow more than one species, we can model the abundance of two species with two variables x and y
- If we fix the total abundance x + y = 1, then we introduce competence
- Average fitness balance the equations

$$\dot{x} = ax - \phi x$$
  
 $\dot{y} = by - \phi y$ 

with  $\phi(t) = ax(t) + by(t)$ .

#### Mutation

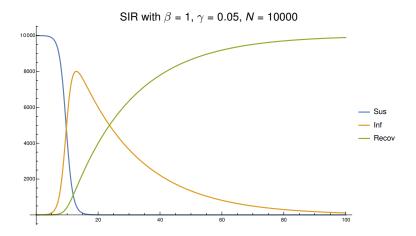
models the rate that one species transform into another one via a mutation

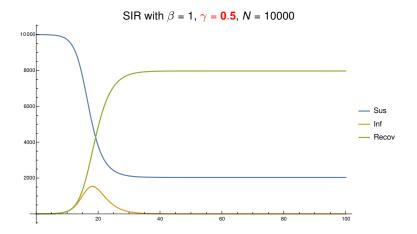
$$\dot{x} = ax - m_{AB}x + m_{BA}y - \phi x$$
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it can be encoded in a mutation matrix Q

Suppose we have an infection in a population of size *N*.

- 1. There are three type of subjects: infected, susceptible, and recovered
- 2. Susceptible agents can be infected
- 3. Infected can infect susceptible
- 4. Recovered cannot infect nor be infected (they are inmune or dead)
- We have the following dynamics
  - 1. An infected agent meets a random agent at rate  $\beta$  and infects her
  - 2. Infected agents recover at rate  $\gamma$
- Can we write the model?





What if we allow the susceptible population to reproduce in our model? e.g. susceptible reproduce at rate  $\alpha$ .

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