

# Influence of a spatial structure on phenotypic evolution

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# Plan

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- 2 Probabilistic model
- 3 Monomorphic populations
- 4 Perspectives

**Aim** : We try to understand the interplay between migration and local competition in the evolution of the phenotypic composition of a population.

First example : heterogeneous environment favors diversity

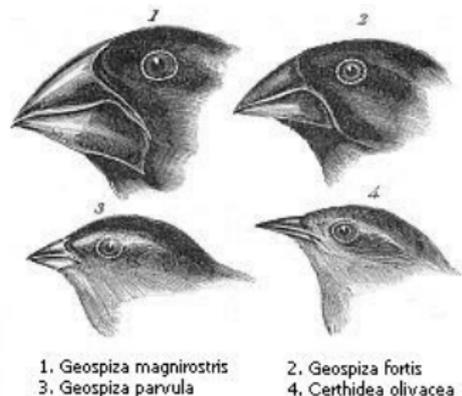


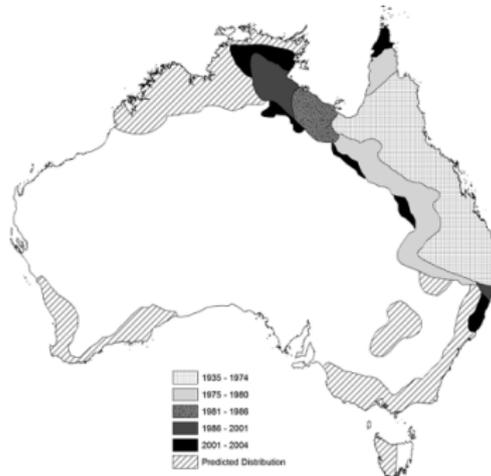
FIGURE : Finches from the Galapagos Archipelago

## Second example : impact of mutation on invasion



**FIGURE :** Female cane toad

Figure 24: Distribution (1935 to 2004) and predicted spread of Cane Toads in Australia



Source: [DEH \(2006\)](#)

**FIGURE :** Distribution (1935 to 2004) and predicted spread of Cane Toads in Australia.  
Source : [www.environment.gov.au](http://www.environment.gov.au)

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- asexual reproduction,
- spatially explicit individual-based model,
- $X_t^i \in \mathcal{X}$  is the location of individual  $i$  at time  $t$ , where  $\mathcal{X}$  is an open bounded subset of  $\mathbb{R}^d$ .
- $U_t^i \in \mathcal{U}$  is the trait of individual  $i$  at time  $t$ , where  $\mathcal{U}$  is a compact set of  $\mathbb{R}^q$ .

## Definition

The population is modeled by the finite measure

$$\nu_t^K = \frac{1}{K} \sum_{i=1}^{N_t} \delta_{(X_t^i, U_t^i)}$$

where  $N_t$  is the number of individuals alive at time  $t$  and  $K > 0$  is a parameter which will be specified later.

# Space evolution

The **migration** of an individual of trait  $u$  is described by a diffusion process normally reflected at the boundary of  $\mathcal{X}$  :

$$\left\{ \begin{array}{l} dX_t = \sqrt{2m(X_t, u)} Id \cdot dB_t + b(X_t, u)dt - dk_t \\ |k|_t = \int_0^t \mathbf{1}_{\{X_s \in \partial\mathcal{X}\}} d|k|_s; \quad k_t = \int_0^t n(X_s) d|k|_s \end{array} \right.$$

where  $k$  is a continuous, increasing process and  $B$  is a  $d$ -dimensional Brownian motion.

## Phenotypic evolution

- **Birth** : An individual described by  $(x, u)$  gives birth to a clonal child at rate  $\lambda_1(x, u)$ ,
- **Death** : An individual described by  $(x, u)$  dies at rate  $\lambda_2(x, u)$ .

We denote its growth rate by  $a(x, u) = \lambda_1(x, u) - \lambda_2(x, u)$ .

- **Birth with mutation** : Each individual gives birth to mutant child at a certain rate, the trait of the child is chosen according to a Gaussian law.

- **Competition** : If the population is described by

$\nu = \frac{1}{K} \sum_{i=1}^n \delta_{(x_i)}$ , the spatial competition against an individual  $(x, u)$  is given by :

$$\mu(x, u) I \star \nu(x, u) = \mu(x, u) \frac{1}{K} \sum_{i=1}^n I(x - x^i)$$

where  $I$  is a competition kernel.

# Simulations

We can then simulate with a computer the behaviour of this population. We present an example here :

- the space  $\mathcal{X}$  is  $(0, 1)$ , and individuals move according to a symmetric diffusion ( $m$  is constant,  $b \equiv 0$ ),
- the space of traits  $\mathcal{U}$  is equal to  $[0, 1]$ ,
- the growth rate is  $a(x, u) = \max(-1, 1 - 20(x - u)^2)$ ,
- two individuals are in competition if and only if their distance is smaller than  $\delta = 0.1$ .

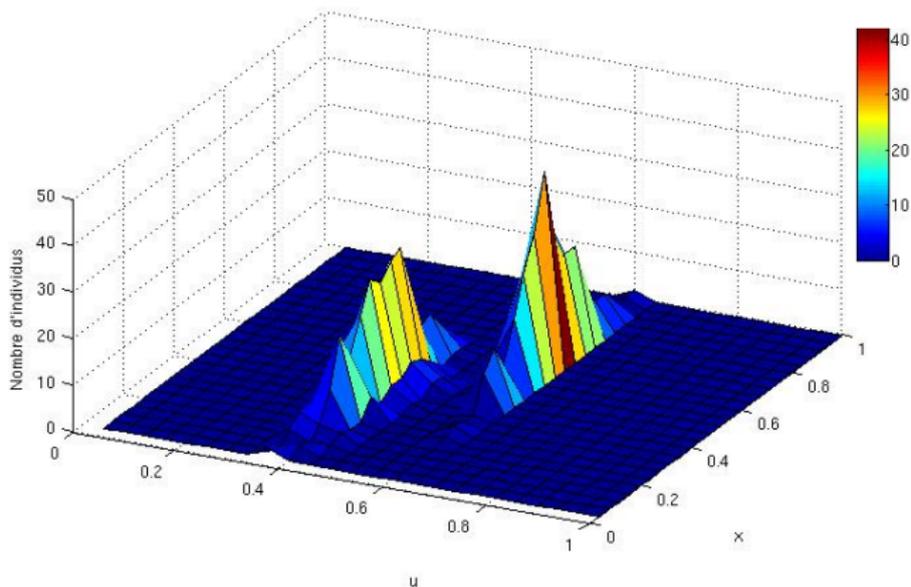


FIGURE : (a)  $t=300$

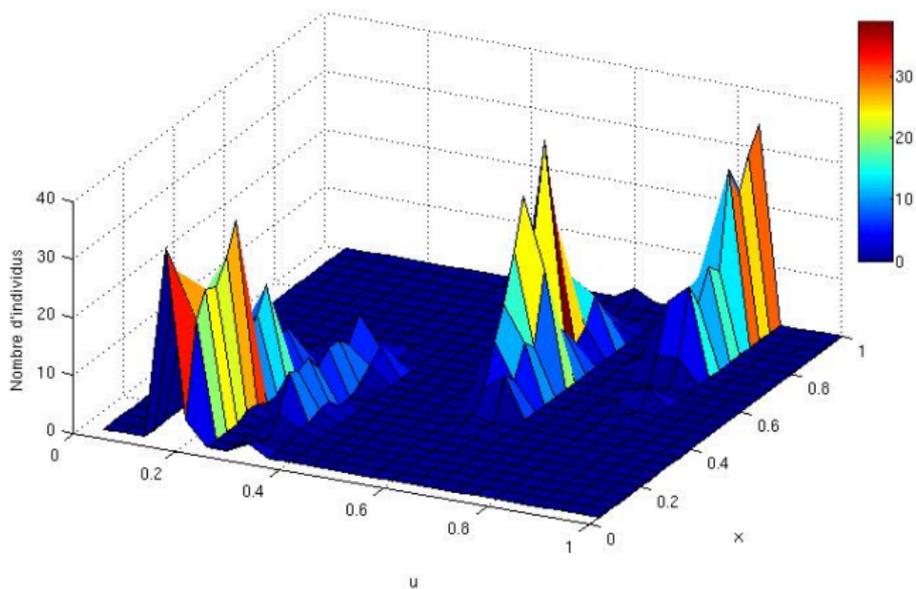


FIGURE : (b)  $t=600$

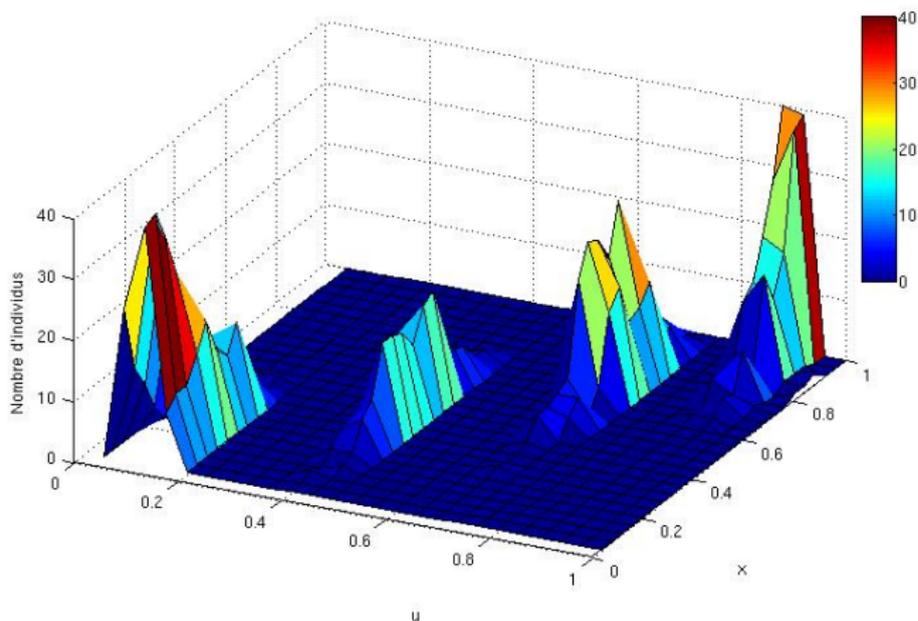


FIGURE : (c)  $t=1500$

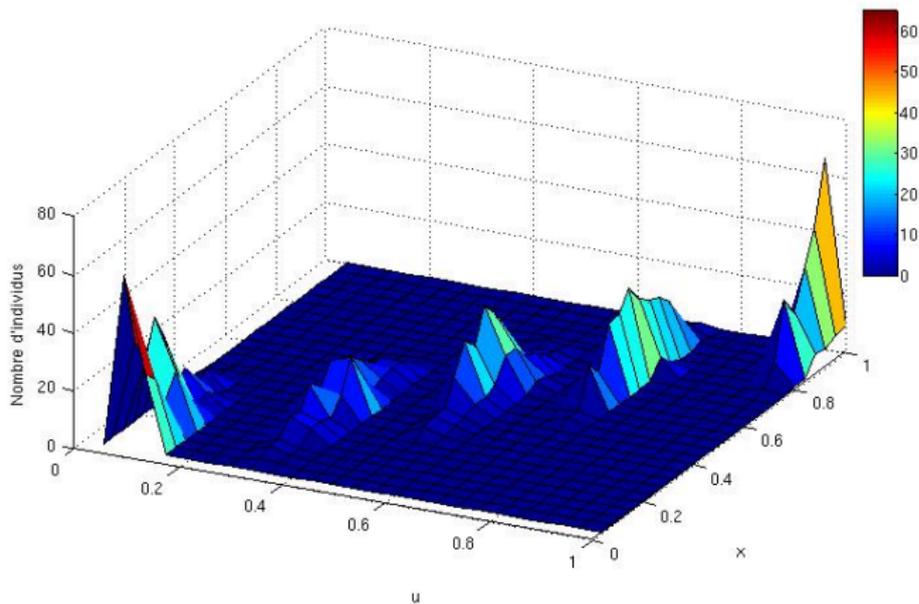


FIGURE : (d)  $t=4000$

Large population approximation :  $K \rightarrow +\infty$

If the coefficients are bounded and if  $m$  has a positive lower bound, the following theorem holds.

### Theorem (Champagnat-Méléard 2007)

*For all  $T > 0$ , if  $(\nu_0^K)_{K>0}$  converges in law to some deterministic finite measure  $\xi_0$  which has a density with respect to Lebesgue measure  $dxdu$  then  $(\nu^K)_{K>0}$  converges in law as a process in  $\mathbb{D}([0, T], M_F(\bar{\mathcal{X}} \times \mathcal{U}))$  to a deterministic function  $\xi \in \mathbb{C}([0, T], M_F(\bar{\mathcal{X}} \times \mathcal{U}))$ . For all  $t$ ,  $\xi_t$  has a density with respect to Lebesgue measure.*

The density function  $g_t(x, u)$  is a weak solution to the partial differential equation :

$$\left\{ \begin{array}{l} \partial_t g_t(x, u) = \Delta(m(x, u)g_t(x, u)) - \nabla(b(x, u)g_t(x, u)) \\ \quad + a(x, u)g_t(x, u) \\ \quad + \mu(x, u) \int_{\mathcal{X}} l(x - y)g_t(y, u)dyg_t(x, u), \\ g_0 \text{ is the density function of } \xi_0, \\ \partial_n g_t(x, u) = 0 \quad \forall (t, x, u) \in [0, T] \times \partial\mathcal{X} \times \mathcal{U}. \end{array} \right.$$

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In the case of a monomorphic population, the evolution equation is

$$\left\{ \begin{array}{l} \partial_t g_t(x) = \Delta(mg_t)(x) + a(x)g_t(x) \\ \quad - \mu(x) \left( \int_{\mathcal{X}} l(x-y)g_t(y)dy \right) g_t(x), \forall x \in \mathcal{X} \\ \partial_n g_t(x) = 0, \forall x \in \partial\mathcal{X}, \forall t \in \mathbb{R}, \end{array} \right.$$

where  $g_t$  is the density at time  $t$  of the population on  $\mathcal{X}$ .

**Result** : if  $l \equiv 1$ ,  $g_t$  tends to a stationary state when  $t$  tends to  $+\infty$ .

## Lemma

If

$$\min_{u \in K^1} \frac{1}{\|u\|_{L^2}^2} \left[ \int_{\mathcal{X}} m |\nabla u|^2 dx - \int_{\mathcal{X}} a(x) u^2(x) dx \right] = -C_a < 0,$$

then

$$\begin{cases} -\Delta(mg)(x) = \left( a(x) - \mu(x) \int_{\mathcal{X}} l(x-y)g(y)dy \right) g(x) \text{ sur } \mathcal{X} \\ \partial_n g(x) = 0 \text{ pour tout } x \in \partial\mathcal{X}. \end{cases}$$

has a positive solution in  $L^2(\mathcal{X})$ .

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**Proof** : Find a **fixed-point** of a function  $\chi$ .

Using the Krein-Rutman theorem,

### Theorem (Krein-Rutman)

Let  $E$  be a Banach space, and  $A : \begin{pmatrix} E & \mapsto & E \\ f & \rightarrow & g \end{pmatrix}$ .

If  $A$  is continuous, compact, and if there exists a closed cone  $K$  such that :

$$\text{if } g \in K, \text{ then } A(g) \in \text{Int}\{K\},$$

then there exists a simple positive eigenvalue of  $A$  with eigenvector in  $\text{Int}\{K\}$ .

we conclude that for all  $f \in L^2$ , there exist  $c_f \in \mathbb{R}$  and  $g \in C^2$ , positive, such that

$$\begin{cases} L_f(g) = c_f g & \forall x \in \mathcal{X} \\ \partial_n g(x) = 0 & \forall x \in \partial\mathcal{X}. \end{cases}$$

where  $L_f(g) = -\Delta(mg) - (a - \mu(l * f))g$ .

We construct a function  $\chi : \begin{pmatrix} L^2(\mathcal{X}) & \mapsto & L^2(\mathcal{X}) \\ f & \rightarrow & g \end{pmatrix}$ ,

$g$  is the positive eigenvector, defined previously, such that

$$c_g = 0.$$

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$$c_g = \min_{\substack{u \in H^1, u > 0, \\ \partial_n u = 0}} \frac{1}{\|u\|_{L^2}^2} \left[ \int_{\mathcal{X}} m |\nabla u|^2 dx - \int_{\mathcal{X}} a(x) u^2(x) dx + \int_{\mathcal{X}} \mu(x) (I * g)(x) u^2(x) dx \right].$$

If  $\chi$  has a fixed point  $g$ , then  $\chi(g) = g$ , so

$$\begin{cases} L_g(g) = c_g g & \forall x \in \mathcal{X} \\ \partial_n g(x) = 0 & \forall x \in \partial\mathcal{X} \end{cases}$$

and  $c_g = 0$ .

i.e.

$$\begin{cases} -\Delta(mg) = (a - \mu(I * g))g & \forall x \in \mathcal{X} \\ \partial_n g(x) = 0 & \forall x \in \partial\mathcal{X} \end{cases}$$

## Fixed-point theorem

We used this fixed point theorem.

### Theorem (Schaefer)

Let  $E$  be a Banach space,  $\chi : E \mapsto E$ , continuous, compact and such that there exists  $R > 0$  satisfying :

*if  $\exists g \in E$ ,  $g = t\chi(g)$ , with  $t \in [0, 1[$ , then  $\|g\|_E \leq R$ ,*

*then  $\chi$  has a fixed point in  $E$ .*

# Smoothness of $\chi$

- $\chi$  is compact :

We take  $A = \{f \in L^2, \|f\|_{L^2} \leq M\}$  and show that  $\chi(A)$  is a bounded set of  $H^1$ , so it is a compact of  $L^2$ .

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## Degree hypothesis

Let  $t$  be in  $]0, 1[$  and  $g$  be in  $L^2$  such that  $g = t\chi(g)$ .

- If  $\mathcal{X} \subset \mathbb{R}$  then

$$\|g\|_{L^2} \leq C\|g\|_{\infty} \leq R.$$

So  $\chi$  has a fixed point, i.e. there exists a positive stationary solution.

- If  $\mathcal{X} \subset \mathbb{R}^d$ ,  $d > 1$ , and if we add the hypothesis

$$(H) \exists k, k' > 0 / \forall x \in \mathcal{X}, k \leq I(x) \leq k',$$

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- Study the evolution of a population with two or more traits.

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Thank you !