

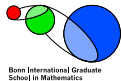
# Stochastic individual-based models of adaptive dynamics

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# Content

- 1 Motivation
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- 3 Some notation and a summary of the relevant known results
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**Adaptive dynamics** aim to study the interplay between **ecology** and **evolution**.

Darwinian evolution of a quantitative trait is the consequence of:

**heredity** **mutation** and **selection**.

The first papers in this context appeared during 1990s. They incorporate the three basic mechanism to describes the evolutionary dynamics.

J. Hofbauer and K. Sigmund. Adaptive dynamics and evolutionary stability. *Appl. Math. Lett.*, 3(4): 75-79, 1990.

P. Marrow, R. Law and C. Cannings. The coevolution of predator-prey interactions: ESSs and Red Queen dynamics. *Proc. R. Soc. Lond. B* 250 : 133-141, 1992.

J.A.J. Metz, R.M. Nisbet and S.A.H. Geritz. How should we define 'fitness' for general ecological scenarios? *Trends Ecol. Evolut.* 7: 198-202, 1992.

## Important advance was the canonical equation of adaptive dynamics

ODE that describes the change of the expected trait value for a population with asexual reproduction, introduced by

U. Dieckmann and R. Law. The dynamical theory of coevolution: a derivation from stochastic ecological processes. *J. Math. Biol.*, 34: 579-612, 1996.

As described in

N. Champagnat, R. Ferrière and G. Ben Arous. The Canonical Equation of Adaptive Dynamics: A Mathematical View. *Selection*, 2: 73-83, 2001.

The critical assumptions in this context are:

- 1 ecological and evolutionary timescales are separated: the population of the wild type is at its ecological equilibrium, when a mutation occurs.
- 2 exclusion principle: a mutant either replaces the wild type or it is eliminated - no long coexistence is allowed

Then, CEAD stems from **three hierarchical scalings**.

- 1 large population size
- 2 very small mutation probability  
(justifies that the ecological and evolutionary timescales are separated)
- 3 very small mutation step

The challenging problems of mathematics are:

- 1 to define an exact "individual-based" model, including the three basis mechanisms and to prove that the involved processes well defined

N. Fournier and S. Méléard. A microscopic probabilistic description of a locally regulated population and macroscopic approximations. *Ann. Appl. Probab.*, 14(4): 1880-1919, 2004.

- 2 to identify in which sense the "real" process converge toward the CEAD. More precisely, to identify the limits that can applied to the "individual-based" model to recover the CEAD.

# The individual-based model

We consider a population with asexual reproduction. Each individual is characterized by a phenotypic trait. The **trait space** is called  $\mathcal{X} \subset \mathbb{R}$  compact. For any two individuals with traits  $x_1, x_2$ ,

- $b(x_1)$  rate of birth
- $d(x_1)$  rate of natural death
- $c(x_1, x_2)$  competition kernel
- $m(x_1)$  probability of mutation in a birth
- $M(x_1, dh)$  mutation law of a mutant trait  $x_1 + h \in \mathcal{X}$ , born from an individual with trait  $x_1$

At any time  $t$ , we consider a finite number  $N_t$  of individuals and denote their traits by  $x_1(t), \dots, x_{N_t}(t)$ .

population state at time  $t$ :  $\tilde{v}_t = \sum_{i=1}^{N_t} \delta_{x_i(t)}$

We want to study the limits of

large population, rare mutations and small mutational effects.

Therefore, we introduce:

- $K$  (large) scaling parameter for **population size**  
 $\Rightarrow \tilde{\nu}_t \rightarrow \nu_t = \frac{1}{K} \tilde{\nu}_t$  and  $c(x, y) \rightarrow \frac{c(x, y)}{K}$
- $\mu$  (small) scaling parameter for **mutation probability**  
 $\Rightarrow m(x) \rightarrow \mu m(x)$
- $\sigma$  (small) scaling parameter for **mutation size**  
 $\Rightarrow$  mutant trait  $x + h \rightarrow x + \sigma h$ , where  $h \sim M(x, dh)$ .

density of  $x$  at time  $t$ :  $\langle \nu_t^{K, \mu, \sigma}, \mathbb{1}_{\{x\}} \rangle$ , where  $\langle \nu, f \rangle := \int_{\mathcal{X}} f d\nu$  and  $\langle \nu_t, \mathbb{1} \rangle = \frac{N_t}{K}$ .

The dynamics of the population can be summarized as follows:

- ① At  $t = 0$ , we have a (possibly random) distribution  $\nu_0^{K, \mu, \sigma}$ .
- ② Each individual (with trait  $x$ ) has three independent exponential clocks:
  - a birth clock with parameter  $b(x)$ ,
  - a natural death clock with parameter  $d(x)$ ,
  - a competition death clock with parameter  $\sum_{i=1}^{N_t} \frac{c(x, x_i(t))}{K}$ .
- ③ If one of the two death clocks of an individual rings, then this one disappears.
- ④ If the birth clock (of an ind. with trait  $x$ ) rings, then it produces an offspring. Its trait is
  - $x$  with prob.  $1 - \mu m(x)$  and
  - $y = x + \sigma h$  with prob.  $\mu m(x)$ , where  $h \sim M(x, dh)$ .



Thus, the process  $(\nu_t^{K,\mu,\sigma})_{t \geq 0}$  is a measure-valued Markov process with state space

$$\mathcal{M}^K(\mathcal{X}) := \left\{ \frac{1}{K} \sum_{i=1}^n \delta_{x_i} : n \geq 0, x_1, \dots, x_n \in \mathcal{X} \right\}$$

and with infinitesimal generator

$$\begin{aligned}
 L^{K,\mu,\sigma} f(\nu) &= \int_{\mathcal{X}} \left( f\left(\nu + \frac{\delta_x}{K}\right) - f(\nu) \right) (1 - \mu m(x)) b(x) K \nu(dx) \\
 &\quad \text{birth without mutation (linear in } \nu) \\
 &+ \int_{\mathcal{X}} \int_{\mathbb{R}} \left( f\left(\nu + \frac{\delta_{x+\sigma h}}{K}\right) - f(\nu) \right) \mu m(x) b(x) M(x, dh) K \nu(dx) \\
 &\quad \text{birth with mutation (linear in } \nu) \\
 &+ \int_{\mathcal{X}} \left( f\left(\nu - \frac{\delta_x}{K}\right) - f(\nu) \right) \left( d(x) + \int_{\mathcal{X}} c(x, y) \nu(dy) \right) K \nu(dx) \\
 &\quad \text{death due to age and competition (non-linear in } \nu)
 \end{aligned}$$

The process  $\nu$  is usually called BPDFL-Process after Bolker, Pacala, Dieckmann and Law.

## Some notation and three limit results

We have to admit always suitable smoothness and boundedness conditions on  $b(\cdot)$ ,  $d(\cdot)$ ,  $c(\cdot, \cdot)$  and  $M(\cdot, dh)$  and a convergence condition for  $\nu_0^{K, \mu, \sigma}$  obtain these results.

**Theorem 1:**  $\mu, \sigma$  fixed and  $K \rightarrow \infty$  *Fournier and Méléard, 2004*

Fix  $\mu$  and  $\sigma$ . Then,  $(\nu_t^K)_{0 \leq t \leq T}$  converges for  $K \rightarrow \infty$  to a deterministic, continuous function  $\xi_t$ . The measure-valued function  $\xi$  is the unique solution of an integrodifferential equation.

**Corollary:** (The special case  $\mu = 0$  and  $\xi_0$  is n-morphic)

If  $\xi_0 = \sum_{i=1}^n z_i(0)\delta_{x_i}$ , then  $\xi_t$  is given by  $\xi_t = \sum_{i=1}^n z_i(t)\delta_{x_i}$ , where  $z_i$  is the solution of the competitive system of Lotka-Volterra equations defined by

$$\dot{z}_i = z_i \left( b(x_i) - d(x_i) - \sum_{j=1}^n c(x_i, x_j) z_j \right), \quad 1 \leq i \leq n.$$

and denoted by  $LV(n, (x_1, \dots, x_n))$ .

## Coexistence and Invasion Fitness:

We say that the traits  $x$  and  $y$  **coexist** if the system  $LV(2, (x, y))$  admits a unique, strictly stable equilibrium  $\bar{z}(x, y)$  with  $\bar{z}_1(x, y) > 0$  and  $\bar{z}_2(x, y) > 0$ .

Strictly stable: Eigenvalues of Jacobian matrix of  $LV(2, (x, y))$  at  $\bar{z}(x, y)$  are all strictly negative.

We define the **invasion fitness** by  $f(y; x) = b(y) - d(y) - c(y, x)\bar{z}(x)$ , where  $\bar{z}(x) = \frac{b(x) - d(x)}{c(x, x)}$  is equilibrium of  $LV(1, x)$ .

Asymmetrically: Initial growth rate of one individual with trait  $y$  which appears in a population of with trait  $x$ , which is close to its equilibrium.

### Proposition: Criterion for coexistence.

There is coexistence in the system  $LV(2, (x, y))$  if and only if

$$f(y, x) > 0 \quad \text{and} \quad f(x, y) > 0.$$

**Theorem 2:**  $\sigma$  fixed and  $(K, \mu) \rightarrow (\infty, 0)$

*Champagnat, 2006*

Fix  $\sigma$  and let  $\nu_0^{K, \mu}$  consist only of individuals with trait  $x$ . Let  $\tau$  be the first time where two individuals in the population have traits that coexist.

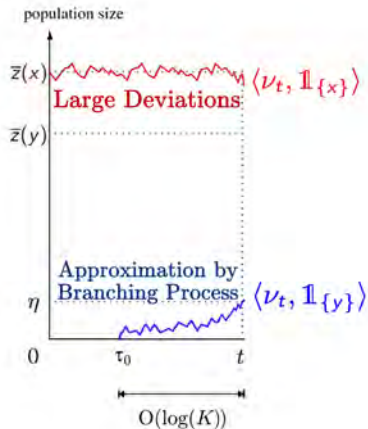
If  $(K, \mu) \rightarrow (\infty, 0)$  in such a way that

$$\forall V > 0, \quad \exp(-VK) \ll \mu \ll \frac{1}{K \log(K)}.$$

Then,  $(\nu_{t/\mu K}^{K, \mu})_{0 \leq t \leq \tau}$  converges to the measure-valued process  $\bar{z}(X_t^\sigma) \delta_{X_t^\sigma}$ ,

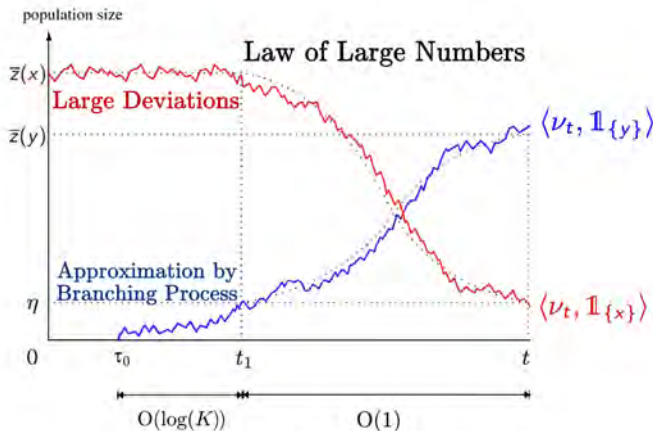
where  $(X_t^\sigma)_{t \geq 0}$  is a stopped Markov jump process with  $X_0^\sigma = x$  known as Trait Substitution Sequence (TSS).

## Sketch of the proof: The invasion of a mutant trait.



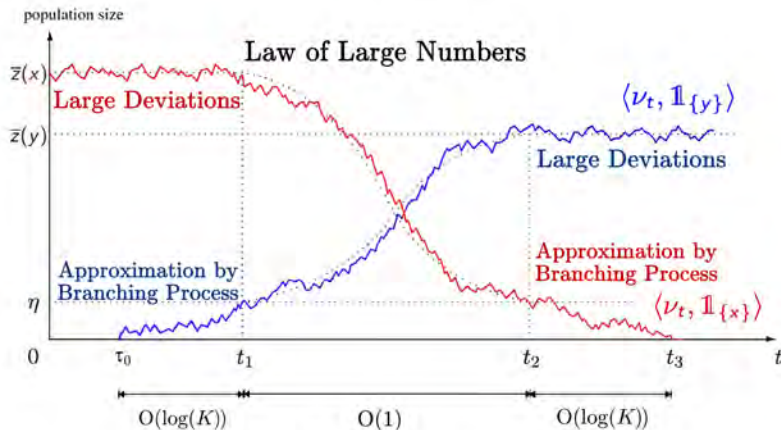
- Step 1:
- $\langle \nu_t^{K,\mu}, \mathbb{1}_{\{x\}} \rangle$  is close to  $\bar{z}(x)$  as long as  $\langle \nu_t^{K,\mu}, \mathbb{1}_{\{y\}} \rangle < \eta$ .
  - $\mathbb{P}[\langle \nu_t^{K,\mu}, \mathbb{1}_{\{y\}} \rangle \text{ reaches } \eta] \rightarrow [f(y, x)]_+ / b(y)$  (probability of non-extinction)

## Sketch of the proof: The invasion of a mutant trait.



- Step 2:
- $\nu_t^{K,\mu}$  is close to the solution of  $LV(2, (x, y))$  with initial state  $(\bar{z}(x), \eta)$ .
  - The solution reaches the  $B_\eta(0, \bar{z}(y))$  in  $O(1)$ .
  - $(0, \bar{z}(y))$  is the equilibrium (not stopped and  $\langle \nu_t^{K,\mu}, \mathbb{1}_{\{y\}} \rangle > \eta$ )

# Sketch of the proof: The invasion of a mutant trait.



- Step 3:
- $\langle \nu_t^{K,\mu}, \mathbb{1}_{\{y\}} \rangle$  stays close to  $\bar{z}(y)$
  - $\langle \nu_t^{K,\mu}, \mathbb{1}_{\{x\}} \rangle$  becomes extinct a.s..



## Sketch of the proof: The invasion of a mutant trait.

The time between two mutations is of order  $\frac{1}{\mu K} \gg \log(K)$ .

Therefore, the three steps of invasion are completed, with high probability, before a new mutation occurs.

So the population process either stops or it is again monomorph, when a new mutation occurs.

**Theorem 3:**  $\sigma \rightarrow 0$  for the TSS.

*Champagnat and Méléard, 2009*

The rescaled Trait Substitution Sequence,  $X_{t/\sigma^2}^\sigma$ , converges for  $\sigma \rightarrow 0$  to the solution of the Canonical Equation of Adaptive Dynamics,  $x(t)$ , where  $x(0) = x$  and

$$\frac{d x(t)}{d t} = \int_{\mathbb{R}} h m(x) \bar{z}(x(t)) [h \partial_1 f(x(t), x(t))]_+ M(x(t), h) d h. \quad (\text{CEAD})$$

The TSS  $X_t^\sigma$  starts with  $X_0^\sigma = x$  and has the generator

$$A\phi(x) = \int_{\mathbb{R}} (\phi(x + \sigma h) - \phi(x)) m(x) b(x) \bar{z}(x) \frac{[f(x + \sigma h, x)]_+}{b(x + \sigma h)} M(x, h) d h,$$

The **jump sizes** are of order  $\sigma$  and the **survival probability** is of order  $\sigma$ .

$\Rightarrow$  one has to rescale the time by  $1/\sigma^2$ .

## Remark:

Theorem 3 does not imply the convergence to the CEAD in any case, when we apply first the limit  $(K, \mu) \rightarrow (\infty, 0)$  and next the limit  $\sigma \rightarrow 0$ .

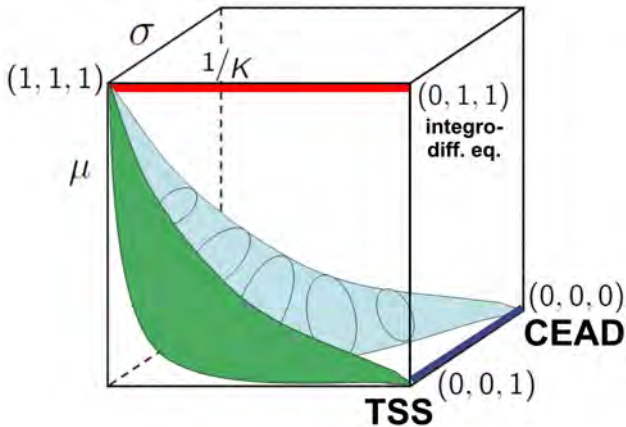
- This is only true if the parameters of the microscopic model prevent the coexistence of any two traits.
- Such a property is called "Invasion Implies Fixation Principle".

## Our goal:

We want to combine the last two limits to a single step i.e. we want to study  $(K, \mu, \sigma) \rightarrow (\infty, 0, 0)$  simultaneously.

We illustrate the different convergences of the model in the following picture.

The convergence of **Theorem 1**, **Theorem 2**, **Theorem 3** and of **our result**.



**Theorem 4:**  $(K, \mu, \sigma) \rightarrow (\infty, 0, 0)$  *B., Bovier, Champagnat, In progress*

Let  $\nu_0^{K, \mu, \sigma}$  consist only of individuals with trait  $x$ .

If  $(K, \mu, \sigma) \rightarrow (\infty, 0, 0)$  in such a way that

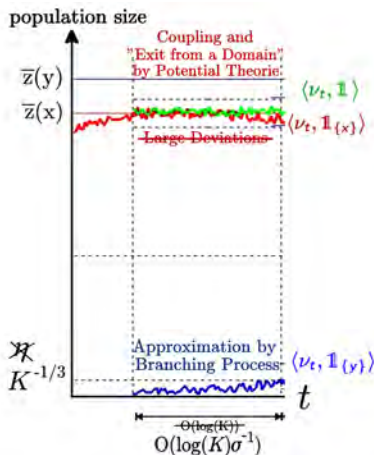
$$K^{-1/3+\alpha} \ll \sigma \ll 1 \quad \text{for some } \alpha > 0$$

$$\text{and } \exp(-CK^{1/3}) \ll \mu \ll \frac{\sigma^3}{K \log K} \quad \text{for all } C > 0,$$

then  $\nu_{t/(\mu K \sigma^2)}^{K, \mu, \sigma}$  converges to the measure-valued function  $\bar{z}(x(t))\delta_{x(t)}$ ,

where  $x(0) = x$  and  $x(t)$  is the unique solution of the Canonical Equation of Adaptive Dynamics (CEAD).

## Sketch of the proof: The three steps of invasion.



- Step 1:
- $\langle \nu_t, \mathbb{1}_{\{x\}} \rangle$  is in  $B_{MK^{-1/3}}(\bar{z}(x))$  while  $\langle \nu_t, \mathbb{1}_{\{y\}} \rangle < K^{-1/3}$ .
  - $\mathbb{P}[\langle \nu_t^{K, \mu, \sigma}, \mathbb{1}_{\{x+\sigma k\}} \rangle \text{ reaches } K^{-1/3}] = \sigma \frac{[k\partial_1 f(x, x)]_+}{b(x)} - o(\sigma)$ .

## Sketch of the proof: The three steps of invasion.

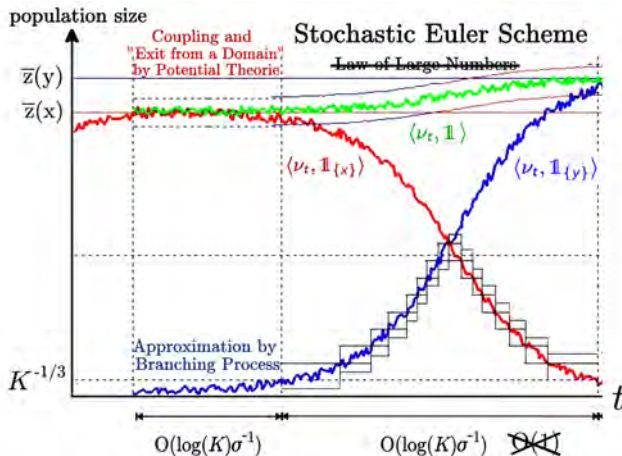
Step 2: Similar as in the Theorem 2?

Problems:

- 1  $\langle \nu_{t_1}^{K, \mu, \sigma}, \mathbb{1}_{\{y\}} \rangle$  is microscopic ( $=K^{-1/3}$ ).
- 2 The fitness advantage of the mutant trait,  $y$ , is only of order  $\sigma$ .  
 $\Rightarrow$  The time  $LV(2, (x, y))$  needs to reach  $B_\epsilon(0, \bar{z}(y))$  is **not**  $O(1)$

Therefore, we can not use the "Law of Large Numbers for Density Depending Population Processes" from Ethier and Kurtz.

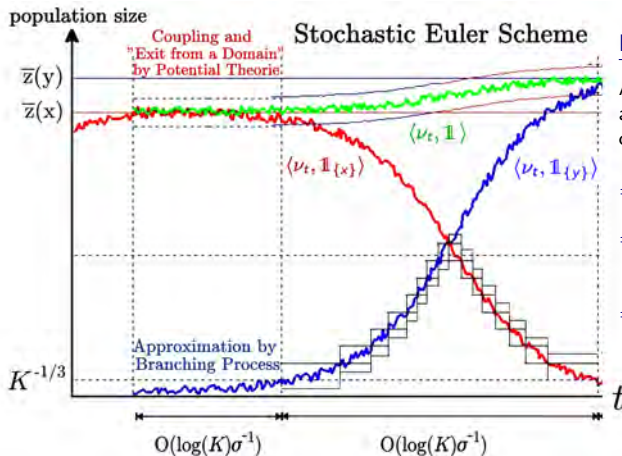
## Step 2: Stochastic Euler Scheme.



Note: If  $\langle \nu^{K, \mu, \sigma}, \mathbb{1}_{\{y\}} \rangle > K^{-1/3}$ , coexistence of the traits  $x$  and  $y$  is not possible.



## Step 2: Stochastic Euler Scheme.



### Frist part:

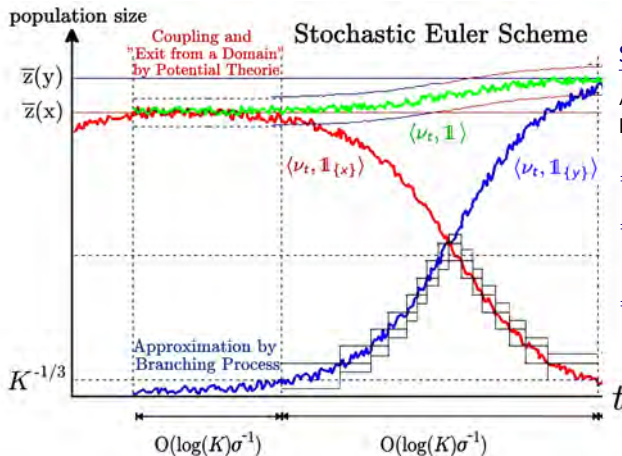
Approximate  $\langle \nu^{K, \mu, \sigma}, \mathbb{1} \rangle$  by a function, which depends only on  $\langle \nu^{K, \mu, \sigma}, \mathbb{1}_{\{y\}} \rangle$ .

$\Rightarrow \langle \nu_t, \mathbb{1}_{\{y\}} \rangle$  drift of order  $+\sigma$

$\Rightarrow$  In little steps prove that  $\langle \nu_t, \mathbb{1}_{\{y\}} \rangle$  increases to  $\bar{z}(x)/2$ .

$\Rightarrow$  Simultaneously  $\langle \nu_t, \mathbb{1}_{\{x\}} \rangle$  decreases to  $\bar{z}(x)/2$ .

## Step 2: Stochastic Euler Scheme.



### Second part:

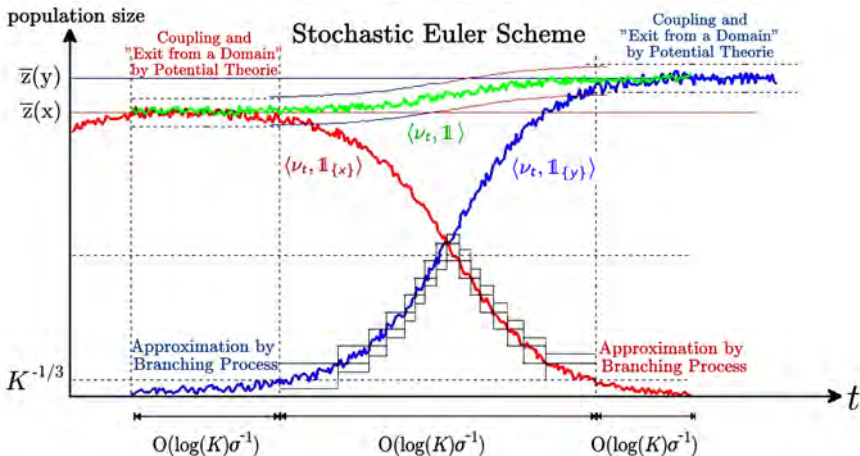
Approximate  $\langle \nu^{K, \mu, \sigma}, \mathbb{1} \rangle$   
by  $\langle \nu^{K, \mu, \sigma}, \mathbb{1}_{\{x\}} \rangle$ .

$\Rightarrow \langle \nu_t, \mathbb{1}_{\{x\}} \rangle$  drift of order  $-\sigma$

$\Rightarrow$  In little steps prove that  
 $\langle \nu_t, \mathbb{1}_{\{x\}} \rangle$  decreases to  $K^{-1/3}$ .

$\Rightarrow$  Simultaneously  $\langle \nu_t, \mathbb{1}_{\{y\}} \rangle$   
increases to  $\bar{z}(y)$ .

# Sketch of the proof: The three steps of invasion.



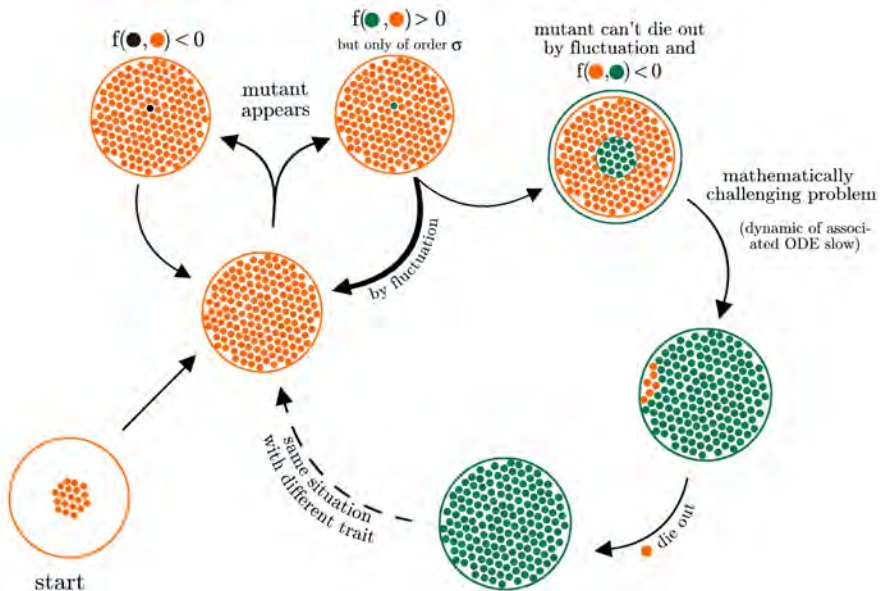
- Step 3:
- $\langle \nu_t^{K,\mu,\sigma}, \mathbb{1}_{\{y\}} \rangle$  stays in  $B_{MK^{-1/3}}\bar{z}(y)$
  - $\langle \nu_t^{K,\mu,\sigma}, \mathbb{1}_{\{x\}} \rangle$  becomes extinct a.s..

## Sketch of the proof: The three steps of invasion.

Since the time between two mutations is of order  $1/(\mu K) \gg \log(K)\sigma^{-1}$ , we obtain with high probability that

- The three invasion steps are completed before a new mutation occurs.  
⇒ The population is again monomorphic, when a new mutation occurs.
- The time intervals where the population is dimorphic convergences to zero for the rescaled process.

## Invasion of a mutant trait



## Sketch of the proof: Approximation by a jump process

Therefore,  $\nu_{t/K\mu\sigma^2}^{K,\mu,\sigma}$  is close to  $\bar{z}(Y_t^\sigma)\delta_{Y_t^\sigma}$ , where  $Y_t^\sigma$  is a Markov jump process with state space  $\mathcal{X}$  and it is constructed as follows:

- it jumps only if a mutant trait invades
- # of mutation until a mutant succeeds is geometrical distributed with an parameter of order  $\sigma$ .
- the jump times of  $Y_t^\sigma$  are again exponentially distributed.

Therefore,  $Y_t^\sigma$  has following generator

$$L^\sigma f(Y) = \int_{\mathbb{R}} (f(Y + \sigma k) - f(Y)) \sigma^{-1} \left( m(Y) \bar{z}(Y) [k \partial_1 f(Y, Y)]_+ \right) M(Y, k) dk$$

But, we can apply the LLN to  $(Y_t^\sigma)_{0 \leq t \leq T}$  s.t. for all  $0 \leq t \leq T$ :

$Y_t^\sigma$  is close to  $x(t)$ , the solution of the CEAD for  $\sigma$  small enough.

Finally, we obtain for  $t < T$ :  $\nu_{t/K\mu\sigma^2}^{K,\mu,\sigma}$  converges to  $\bar{z}(x(t))\delta_{x(t)}$ .

## Summary and Outlook

- have studied the limits  $(K, \mu, \sigma) \rightarrow (\infty, 0, 0)$  in a single step
- interesting to consider other time scales, where coexistence is possible  
 $\Rightarrow$  phenomenon of evolutionary branchings appears

N. Champagnat, S. Méléard, Polymorphic evolution sequence and evolutionary branching. *Prob. Th. Rel. Fields.* 151(1-2): 45-94, 2011.

- interesting to expand the model to one with a fast and a slow timescale for trait changes, i.e. migration and mutation

A. Bovier, S.-D. Wang, Trait substitution trees on two time scales. *Markov Proc. Rel. Fields.* 19, 607-642, 2013.

## An application or extension of the model

This will be joint work with A. Bovier, L. Coquille, H. Mayer and B. Prochnau.

Landsberg J. et al. Melanomas resist T-cell therapy through inflammation-induced reversible dedifferentiation. *Nature* 490(7420):412-6, 2012

M. Hölzel, A. Bovier and T. Tüting, Plasticity of tumor and immune cells: a source of heterogeneity and a cause for therapy resistance? *Nat. Rev. Cancer* 13: 365-376, 2013.

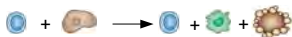
- Melanomas are tumors associated with skin cancer.
- There are several therapeutic approaches in treat cancer cells.  
⇒ One way is to use cytotoxic T-cells which are specific for the melanoma cells and kill these cells.



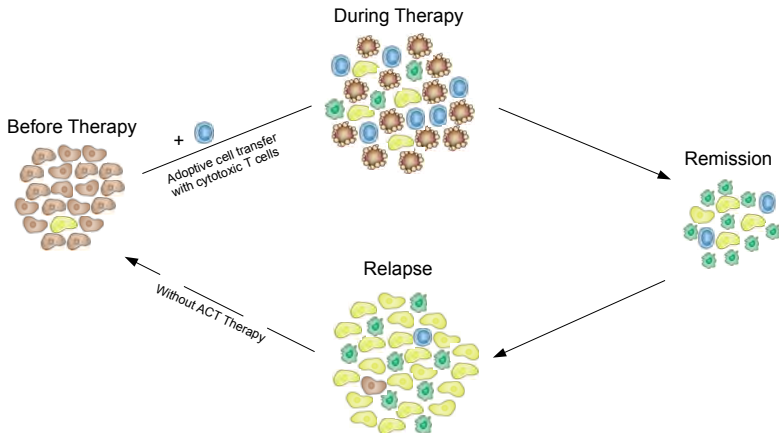
The following problem is reported in the first article:

- During the treatment occurs an inflammation and the cancer cells react to the inflammation by switching their phenotype.
- More precisely, the switch occurs if so called TNF-alpha (Tumor necrosis factor) proteins are present.
- The T-cells are not able to attack this cancer cells with the new phenotype of and thus very often a relapse occurs.
- Furthermore, Landsberg et al. observed that this switch is reversible, i.e. the melanoma cell population switches back again in the absence of TNF-alpha molecules.

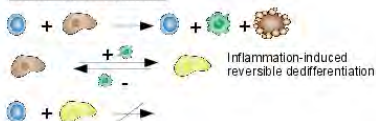
Relevant mechanism



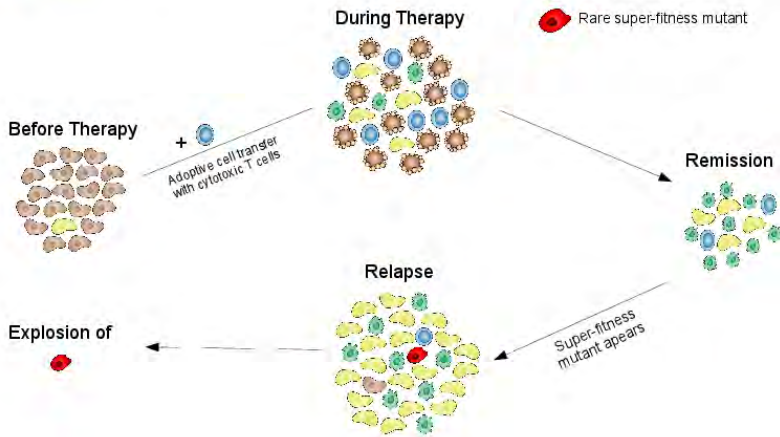
- Tumour cell
- Phenotype-switched tumour cell
- Cytotoxic T cell
- Inflammatory indicator TNF-alpha
- Dying tumour cell



Relevant mechanism



- Tumour cell
- Phenotype-switched tumour cell
- Cytotoxic T cell
- Inflammatory indicator TNF-alpha
- Dying tumour cell
- Rare super-fitness mutant



Goals: Investigate a model for the whole system and numerical simulations.

More precisely, a system which describes the dynamics of

- ① TNF-alpha proteins
- ② cytotoxic T-cells, perhaps more than one type and
- ③ melanoma cells with different phenotype and also with different genotype to be able to include superfit mutants

## What will change in the model?

- 1 have to include an extra death rate due to cytotoxic T-cells treatment for some melanoma cells .
- 2 have to include natural switching rates between the different phenotypes for the melanoma cells with the same genotype and some due to inflammation.
- 3 like to include a competition which lowers the birth rate  
⇒ increases probability of a mutations in a small tumor.
- 4 have to take care of the dynamics of the TNF-alpha protein and the cytotoxic T-cells

**Thank you for your attention!**