

# Speciation as Pattern Formation by Competition in a Smooth Fitness Landscape

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## Abstract

We investigate the problem of speciation and coexistence in simple ecosystems when the competition among individuals is included in the Eigen model for quasi-species. By suggesting an analogy between the competition among strains and the diffusion of a chemical inhibitor in a reaction-diffusion system, the speciation phenomenon is considered the analogous of chemical pattern formation in genetic space. In the limit of vanishing mutation rate we obtain analytically the conditions for speciation. Using different forms of the competition interaction we show that the speciation is absent for the genetic equivalent of a normal diffusing inhibitor, and is present for shorter-range interactions. The comparison with numerical simulations is very good.

In this work we address the problem of speciation (species formation) in simple ecosystems, mirroring some aspects of bacterial and viral evolution. Our model can be considered as an extension of the Eigen model [1,2]. With respect to the latter, we introduce the competition among individuals.

Eigen's phenomenological theory of self-reproducing macromolecules (or haploid organisms) illustrates the concept of stable quasi-species, i.e. a peaked distribution of genomes around a master sequence, its width being determined by mutations. In its simpler formulation, the various genomes have different reproductive rates, the logarithm of which constitutes the fitness landscape [3–5]. The master sequence is located in correspondence of the maximum of the fitness. In general a one to one correspondence between a given phenotype and a genotype is assumed (no polymorphism nor age structure). The genomes are coupled by mutations and by a global constraint on the total number of individuals (constant organization). One usually considers only point mutations (the most common ones), which correspond to a diffusion process in genetic space. In this way one can define the concept of distance in genetic space as the number of mutations needed to connect two genomes. The Eigen model has also been studied in the context of statistical mechanics [6–9].

For a vanishing mutation rate (which is the case for actual organisms), only one quasi-species can survive in the long time limit, except for the marginal case of degenerate maxima of the fitness or for an extremely rough landscape (similar to the spin glass energy landscape) for which the system never attains equilibrium. Epstein [10] studied the problem without considering mutations; he showed that the coexistence is possible if the species are self-limiting (i.e. there exists a form of self-competition, modeled for instance by a logistic term) and coexisting species does not compete directly. On the contrary, when two species are in competition (because they share some resource — an enzyme in Epstein's case), only the fittest one survives. However, he did not introduce the genetic distance among species nor presented any evolutionary mechanism for the speciation phenomenon.

We think that the direct competition for local resources among strains, coupled with

a weak mutation rate, is the simplest mechanism for modeling both speciation and stable coexistence in simple smooth landscapes. The mutations are needed to populate newly formed niches, while the competition actively separates the strains into quasi-species. One can consider the following analogy with a Turing mechanism for chemical pattern formation. The main ingredients are an autocatalytic reaction process (reproduction) with slow diffusion (mutations) coupled with the emission of a short-lived, fast-diffusing inhibitor (competition). In this way a local high concentration of autocatalytic reactants inhibits the growth in its neighborhood, acting as a local negative interaction.

In genetic space, the local coupling is given by the competition among genetically kin individuals. For instance, assuming a certain distribution of some resource (such as some essential metabolic component for a bacterial population), then the more genetically similar two individuals are, the wider the fraction of shared resources is. The effects of competition on strain  $x$  by strain  $y$  are modeled by a term proportional to the relative abundance of the latter,  $p(y)$ , modulated by a function that decreases with the genetic distance between  $x$  and  $y$ . Another example of this kind of competition can be found in the immune response in mammals. Since the immune response has a certain degree of specificity, a viral strain  $x$  can suffer from the response triggered by strain  $y$  if they are sufficiently near in an appropriate genetic subspace. Again, one can think that this effective competition can be modeled by a term, proportional to the relative abundance of the strain that originated the response, which decreases with the genetic distance.

Although Eigen's model is defined in an hypercubic genetic space, and the error threshold transition rigorously exists only in an infinite-dimensional space [9], the proposed speciation phenomenon is independent on the dimension of the genetic space. We shall work therefore in a linear genetic space. An instance of a similar (sub-)space in real organisms is given by a repeated gene (say a tRNA gene): a fraction of its copies can mutate, linearly varying the fitness of the individual with the "chemical composition" of the gene [11]. This degenerate case has been widely studied (see for instance Refs. [12]); one should introduce the multiplicity of a degenerate state, which can be approximated to a Gaussian, but if one works in

the neighborhood of its maximum (the most common chemical composition) the multiplicity factors are nearly constants. Another example is given by the level of catalytic activity of a protein. A linear space has also been used for modeling the evolution of RNA viruses on HeLa cultures [13].

Let us start with a one dimensional “chemical” model of cells that reproduce asexually and slowly diffuse (in real space),  $p = p(x, t)$  being their relative abundance at position  $x$  and at time  $t$ . These cells constitutively emit a short-lived, fast-diffusing mitosis inhibitor  $q = q(x, t)$ . This inhibitor may be simply identified with some waste or with the consumption of a local resource (say oxygen). The diffusion of the inhibitor is modeled as

$$\frac{\partial q}{\partial t} = k_0 p + D \frac{\partial^2 q}{\partial x^2} - k_1 q, \quad (1)$$

where  $k_0$ ,  $k_1$  and  $D$  are the production, annihilation and diffusion rates of  $q$ .

The evolution of the distribution  $p$  is given by

$$\frac{\partial p}{\partial t} = \left( A(x, t) - \bar{A}(t) \right) p + \mu \frac{\partial^2 p}{\partial x^2}, \quad (2)$$

$$\bar{A}(t) = \int A(y, t) p(y, t) dy. \quad (3)$$

The growth rate  $A$  can be expressed in terms of the fitness  $H$  as

$$A(x, t) = \exp(H(x, t)). \quad (4)$$

Due to the form of equation (2), the distribution  $p$  is always normalized to one. The diffusion rate of  $q$ ,  $D$ , is assumed to be much larger than  $\mu$ . The growth rate  $A$ , can be decomposed in two factors,  $A(x, t) = A_0(x)A_1(q(x, t))$ , where  $A_0$  gives the reproductive rate in absence of  $q$ , so  $A_1(0) = 1$ . In presence of a large concentration of the inhibitor  $q$  the reproduction stops, so  $A_1(\infty) = 0$ . A possible choice is

$$A(x, t) = \exp(H_0(x) - q(x, t)).$$

For instance,  $H_0(x)$  could model the sources of food or, for algae culture, the distribution of light.

Since we assumed a strong separation in time scales, we look for a stationary distribution  $\tilde{q}(x, t)$  of the inhibitor (eq. (1)) by keeping  $p$  fixed. This is given by a convolution of the distribution  $p$ :

$$\tilde{q}(x, t) = J \int \exp\left(-\frac{|x-y|}{R}\right) p(y, t) dy,$$

where  $J$  and  $R$  depend on the parameters  $k_0, k_1, D$ . In the following we shall use  $J$  and  $R$  as control parameters, disregarding their origin.

We can generalize this scenario to non-linear diffusion processes of the inhibitor by using the reaction-diffusion equation eq. (2), with the fitness  $H$  and the kernel  $K$  are given by

$$H(x, t) = H_0(x) - J \int K\left(\frac{x-y}{R}\right) p(y, t) dy \quad (5)$$

$$K(r) = \exp\left(-\frac{|r|^\alpha}{\alpha}\right), \quad (6)$$

i.e. a symmetric decreasing function of  $r$  with  $K(0) = 1$ . The parameters  $J$  and  $\alpha$  control the intensity of the competition and the steepness of the interaction, respectively.

Let us consider the correspondence with the genetic space: the quantity  $x$  now identifies a genome, the diffusion rate  $\mu$  is given by mutations, and the inhibitor  $q$  (which is no more a real substance) represents the competition among genetically related strains. The effects of competition are much faster than the genetic drift (mutations), so that the previous hypotheses are valid. While the genetic interaction kernel  $K(r)$  is not given by a diffusion process, its general form should be similar to that of eq. (6): a decreasing function of the genetic distance between two strains. We shall refer to the  $p$ -independent contribution to the fitness,  $H_0(x)$ , as the static fitness landscape.

Our model is thus defined by eqs. (2–6). We are interested in its asymptotic behavior in the limit  $\mu \rightarrow 0$ . Actually, the mutation mechanism is needed only to define the genetic distance and to allow population of an eventual niche. The results should not change qualitatively if one includes more realistic mutation mechanisms.

Let us first examine the behavior of eq. (2) in absence of competition ( $J = 0$ ) for a smooth static landscape and a vanishing mutation rate. This corresponds to the Eigen

model in one dimension: since it does not exhibit any phase transition, the asymptotic distribution is unique. The asymptotic distribution is given by one delta function peaked around the global maximum of the static landscape, or more delta functions (coexistence) if the global maxima are degenerate. The effect of a small mutation rate is simply that of broadening the distribution from a delta peak to a bell-shaped curve [14].

While the degeneracy of maxima of the static fitness landscape is a very particular condition, we shall show in the following that in presence of competition this is a generic case. For illustration, we report in Figure 1 the numerical computation of the asymptotic behavior of the model for a possible evolutive scenario that leads to the coexistence of three species. We have chosen a smooth static fitness  $H_0$  (see eq. (7)) and a Gaussian ( $\alpha = 2$ ) competition kernel. The effective fitness  $H$  is almost degenerate (here  $\mu > 0$  and the competition effect extends on the neighborhood of the maxima), and this leads to the coexistence. One could show that the curvature of the maxima affects the width and the height of the quasi-species distribution in presence of mutations [14].

We shall now derive the conditions for the coexistence of multiple species. Let us assume that the asymptotic distribution is formed by  $L$  delta peaks  $p_k$ ,  $k = 0, \dots, L - 1$ , for a vanishing mutation rate (or  $L$  non-overlapping bell shaped curves for a small mutation rate) centered at  $y_k$ . The weight of each quasi species is  $\gamma_k$ , i.e.

$$\int p_k(x)dx = \gamma_k, \quad \sum_{k=0}^{L-1} \gamma_k = 1.$$

The quasi-species are ordered such as  $\gamma_0 \geq \gamma_1, \dots, \geq \gamma_{L-1}$ .

The evolution equations for the  $p_k$  are ( $\mu \rightarrow 0$ )

$$\frac{\partial p_k}{\partial t} = (A(y_k) - \bar{A})p_k,$$

where  $A(x) = \exp(H(x))$  and

$$H(x) = H_0(x) - J \sum_{j=0}^{L-1} K\left(\frac{x - y_j}{R}\right) \gamma_j.$$

The stability condition of the asymptotic distribution is  $(A(y_k) - \bar{A})p_k = 0$ , i.e. either  $A(y_k) = \bar{A} = \text{const}$  (degeneracy of maxima) or  $p_k = 0$  (all other points). In other terms one

can say that in a stable environment the fitness of all individuals is the same, independently on the species.

The position  $y_k$  and the weight  $\gamma_k$  of the quasi-species are given by  $A(y_k) = \bar{A} = \text{const}$  and  $\partial A(x)/\partial x|_{y_k} = 0$ , or, in terms of the fitness  $H$ , by

$$H_0(y_k) - J \sum_{j=0}^{L-1} K\left(\frac{y_k - y_j}{R}\right) \gamma_j = \text{const}$$

$$H'_0(y_k) - \frac{J}{R} \sum_{j=0}^{L-1} K'\left(\frac{y_k - y_j}{R}\right) \gamma_j = 0$$

Let us compute the phase boundary for coexistence of three species for two kinds of kernels: the exponential (diffusion) one ( $\alpha = 1$ ) and a Gaussian one ( $\alpha = 2$ ).

We assume that the static fitness  $H_0(x)$  is a symmetric linear decreasing function except in the vicinity of  $x = 0$ , where it has a quadratic maximum:

$$H_0(x) = b \left( 1 - \frac{|x|}{r} - \frac{1}{1 + |x|/r} \right) \quad (7)$$

so that close to  $x = 0$  one has  $H_0(x) \simeq -bx^2/r^2$  and for  $x \rightarrow \infty$ ,  $H_0(x) \simeq b(1 - |x|/r)$ . We have checked numerically that the results are qualitatively independent on the exact form of the static fitness, providing that it is a smooth decreasing function.

Due to the symmetries of the problem, we have one quasi-species at  $x = 0$  and two symmetric quasi-species at  $x = \pm y$ . Neglecting the mutual influence of the two marginal quasi-species, and considering that  $H'_0(0) = K'(0) = 0$ ,  $K'(y/R) = -K'(-y/r)$ ,  $K(0) = J$  and that the three-species threshold is given by  $\gamma_0 = 1$  and  $\gamma_1 = 0$ , we have

$$\tilde{b} \left( 1 - \frac{\tilde{y}}{\tilde{r}} \right) - K(\tilde{y}) = -1,$$

$$\frac{\tilde{b}}{\tilde{r}} + K'(\tilde{y}) = 0.$$

where  $\tilde{y} = y/R$ ,  $\tilde{r} = r/R$  and  $\tilde{b} = b/J$ . In the following we drop the tildes for convenience.

Thus

$$r - z - G \exp\left(-\frac{z^\alpha}{\alpha}\right) = -G,$$

$$Gz^{\alpha-1} \exp\left(-\frac{z^\alpha}{\alpha}\right) = 1,$$

where  $G = r/b$ .

For  $\alpha = 1$  we have the coexistence condition

$$\ln(G) = r - 1 + G.$$

The only parameters that satisfy these equations are  $G = 1$  and  $r = 0$ , i.e. a flat landscape ( $b = 0$ ) with infinite range interaction ( $R = \infty$ ). Since the coexistence region reduces to a single point, it is suggested that  $\alpha = 1$  is a marginal case.

For  $\alpha = 2$  the coexistence condition is given by

$$z^2 - (G + r)z + 1 = 0,$$

$$Gz \exp\left(-\frac{z^2}{2}\right) = 1.$$

One can solve numerically this system and obtain the boundary  $G_c(r)$  for the coexistence. In the limit  $r \rightarrow 0$  (static fitness almost flat) one has

$$G_c(r) \simeq G_c(0) - r \tag{8}$$

with  $G_c(0) = 2.216\dots$

We have performed several numerical simulations for different values of the parameters, whose results are presented in Figure 2. The boundary of the multi-species phase is well approximated by eq. (8); in particular, this boundary does not depend on the mutation rate  $\mu$ , at least for  $\mu < 0.1$ , which can be considered a very high mutation rate for real organisms. The most important effect of  $\mu$  is the broadening of quasi-species curves, which can eventually merge.

In conclusion, we have introduced a model for the genetic evolution of haploid organisms under the pressure of a static fitness landscape and competition. This model exhibits the



phenomenon of species formation in a way reminiscent of a chemical pattern formation via a Turing-like mechanism. We have analyzed analytically this system in the limit of vanishing mutation rate and linear genetic space, showing that an increasing level of a short-range competition induces a transition from a single species distribution to a stable environment in which multiple genetically distinct species are present. The comparison of the analytical approximation with the numerical integration of the original differential equations is very good. We think that the mechanism that we proposed is the simplest one for modeling speciation and species coexistence in a smooth (or flat) fitness landscape.

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## FIGURE CAPTIONS

Fig. 1. Static fitness  $H_0$ , effective fitness  $H$ , and asymptotic distribution  $p$  numerically computed for the following values of parameters:  $\alpha = 2$ ,  $\mu = 0.01$ ,  $H_0 = 1.0$ ,  $b = 0.04$ ,  $J = 0.6$ ,  $R = 10$ ,  $r = 3$  and  $N = 100$ .

Fig. 2. Three-species coexistence boundary  $G_c$  for  $\alpha = 2$ . The continuous line represents the analytical approximation, eq. (8), the circles are obtained from numerical simulations. The error bars represent the maximum error.

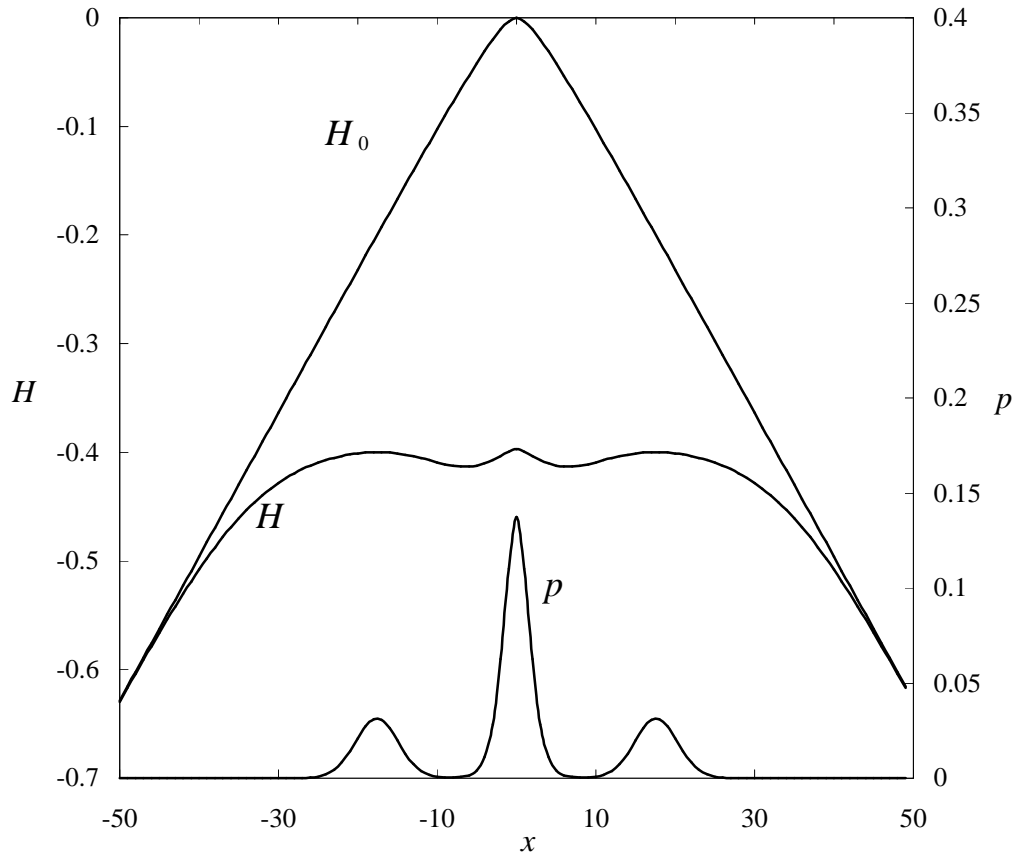


figure 1

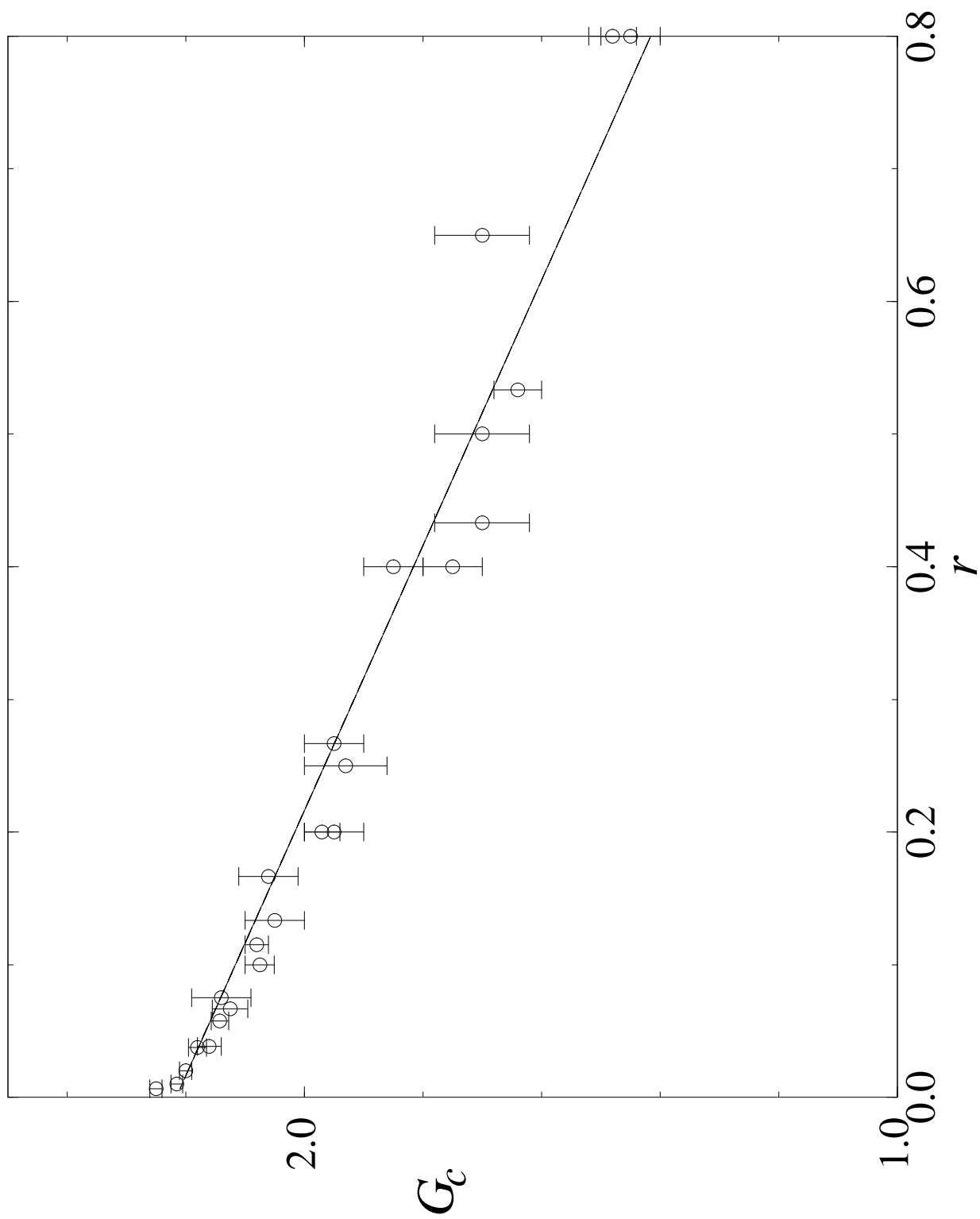


Figure 2