Sympatric speciation in an age-structured population living on a lattice

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Abstract. A square lattice is introduced into the Penna model for biological aging in order to study the evolution of diploid sexual populations under certain conditions when one single locus in the individual’s genome is considered as identifier of species. The simulation results show, after several generations, the flourishing and coexistence of two separate species in the same environment, i.e., one original species splits up into two on the same territory (sympatric speciation). As well, the mortalities obtained are in a good agreement with the Gompertz law of exponential increase of mortality with age.

1 Introduction

The understanding of species formation - groups of actually or potentially inter-breeding populations, which are reproductively isolated from other such groups - is still a fundamental problem in biology [1]. Speciation usually occurs when a pre-existing population is divided into two or more smaller populations by a geographical barrier, like an island, river, isolated valley, or mountain range. Once reproductively isolated by the barrier, the gene pools in the two populations can diverge due to natural selection, genetic drift, or gene flow, and if they sufficiently diverge, the inter-breeding between the populations will not occur if the barrier is removed. As a result, new species have been formed.

In spite of theoretical difficulties to show convincingly how speciation might occur without physical separation [2], there is an increasing evidence for the process of sympatric speciation, in which the origin of two or more species from a single ancestral one occurs without geographical isolation [3]. The most straightforward scenario for sympatric speciation requires disruptive selection favoring two substantially different phenotypes, followed by the elimination of all intermediate phenotypes. In sexual populations, the stumbling block preventing sympatric speciation is that mating between divergent ecotypes constantly scuffles gene combinations, creating organisms with intermediate phenotypes. However, this mixing can be prevented if there is assortative [4] instead of random mating, i.e., mating of individuals that are phenotypically similar. It can be based on ecologically important traits such as body size (as in stickle-backs) [5] or on marker traits that co-vary with ecological traits (such as coloration or breeding behavior in cichlids) [6].

The present paper reports on an attempt to address the challenging problem of sympatric speciation using the widespread Penna bit-string model [7] for age-structured populations, which is based on the mutation accumulation theory for biological aging. It has successfully reproduced many different characteristics of living species, as the catastrophic senescence of pacific salmon [9], the inheritance of longevity [10] and the evolutionary advantages of sexual reproduction [11], as well as a particular case of sympatric speciation [12].

2 The model

2.1 Model without lattice

Each individual of the population is represented by a “chronological genome”, which consists of two bit-strings of 32 bits (32 loci or positions) each, that are read in parallel. One string contains the genetic information inherited from the mother and the other, from the father. Each position of the bit-strings is associated to a period of the individual’s life, which means that each individual can live at most for 32 periods (“years”). Each step of the simulation corresponds to reading one new position of all individuals’ genomes. Genetic defects are represented by bits 1. If an individual has two bits 1 at the \( i \)-th position of both bit-strings (homozygote), it will start to suffer the effects of a genetic disease at its \( i \)-th year of life. If the individual is homozygous with two bits zero, no disease appears
in that age. If the individual is heterozygous in that position, it will become sick only if that locus is one for which the harmful allele is dominant. The dominant loci are randomly chosen at the beginning of the simulation and remain fixed. If the current number of accumulated diseases reaches a threshold \( T \), the individual dies.

If a female succeeds in surviving until the minimum reproduction age \( R \), it generates \( b \) offspring every year until death. The female randomly chooses a male to mate, the age of which must also be greater or equal to \( R \). The offspring’s genome is constructed from the parent’s ones: first the strings of the mother are randomly crossed, and a female gamete is produced. \( M_m \) deleterious mutations are then randomly introduced. The same process occurs with the father’s genome (with \( M_f \) mutations), and the union of the two remaining gametes form the new genome. This procedure is repeated for each of the \( b \) offspring. The sex of the baby is randomly chosen, each one with probability 50%. Deleterious mutation means that if a bit 0 is randomly chosen in the parent’s genome, it is set to 1 in the offspring genome. However, if a bit already set to 1 is randomly chosen, it remains 1 in the offspring genome (no back mutations).

The description given above corresponds to the original sexual version of the Penna model \([13]\), in which at every time step each individual of the population, independently of its age or current number of accumulated diseases, can be killed with a probability \( V_i = 1 - N_i/N_{\text{max}} \); \( N_{\text{max}} \) is the maximum population size (the carrying capacity of the environment) and \( N_i \) is the current population size. This random time-dependent death, well known as the Verhulst factor, is introduced in order to avoid the unlimited growth of the population and to take into account the dispute for food and space. Since there seems to be no biological justification for considering random deaths in real populations, as well as a controversial importance of its role in the Penna model \([14]\), in our simulations we do not consider random deaths. Instead, we adopt a simple lattice dynamics which also avoids the exponential increase of the population. The details will be presented in the next subsection.

\subsection*{2.2 Speciation model on a lattice}

In the present case each individual lives on a given site \((i, j)\) of a square lattice and, at every time-step, has a probability \( p_w \) to move to the neighboring site that presents the smallest occupation, if this occupation is also smaller or equal to that of the current individual’s site. We start the simulations randomly distributing one individual per site on a diluted square lattice. That is, if an already occupied site is chosen for a new individual, the choice is disregarded and another random site is picked out.

At any bit position a diploid individual can have \( n = 0, 1 \) or 2 bits set. The process of sympatric speciation is now attempted by defining one single bit position, which we take as position 11, as an identifier of the species. Mating occurs only among individuals of the same species (same value of \( n \) at position 11), which means that this locus also defines the mating preferences. Each able female (with age \( \geq R \)) with \( n \) such bits randomly selects a neighboring able male with the same \( n \) value to breed. If she succeeds, she generates \( b \) offspring. Then she chooses at random, again among its four neighboring sites, a place to put each baby, according to the rules below. The newborn dies if it is not possible to find a site respecting these rules:

1) The selected site occupation must be \( \leq 1 \);
2) If the newborn has \( n = 0 \), then it can occupy an empty site or a site already occupied by a single individual with \( n = 2 \);
3) If the newborn has \( n = 2 \), then it can occupy an empty site or a site already occupied by a single individual with \( n = 0 \);
4) If the newborn has \( n = 1 \), it can occupy only an empty site.

Rules 2 and 3 mean that the \( n = 0 \) and the \( n = 2 \) populations can share the same habitat, that is, they do not dispute for the same food resources. Rule 4 means that the \( n = 1 \) population feeds at both niches, competing with the other two. Theses rules replace the random killing Verhulst factor pointed out in the previous section.

We start our simulations only with \( n = 0 \) individuals. Due to the randomness of mutations and crossover, the offspring does not necessarily have the same \( n \) value of the parents. In our model it is exactly this randomness which allows the emergence of new species out of the original one. These populations coexist in a stable equilibrium but without cross-mating.

\section*{3 Simulation Results}

The simulation starts with \( N_0 \) individuals, half males and half females, and runs for a pre-specified number of time steps, at the end of which averages are taken over the population(s). The general parameters of the simulations are:

- Minimum age of reproduction \( R = 8 \);
- Birth rate \( b = 3 \);
- Mutation rate \( M = 1 \) per bit-string (or gamete);
- Maximum number of genetic diseases \( T = 5 \);
- Probability to walk \( p_w = 1.0 \).

Fig. \ref{fig:speciation} shows how the new species \( N_2 \) emerges, within about a hundred iterations, from the original species \( N_0 \). The intermediate population \( N_1 \) is only about 0.50\% of the total population. Since the rule for an individual to move on the lattice depends only on the existence of a site with an occupation smaller or equal to that of the current individual’s site and is completely non-related to the individual species, the different species may bunch together at the same site of the lattice. Then, due to the reproduction rules quoted above, after several generations we obtain a great predominance of the two non-competing species \( N_0 \) and \( N_2 \) living at the same geographic position (sympatric speciation). Our results with \( N_0 = 1600 \) are confirmed by larger simulations with \( N_0 = 100000 \), and also by larger simulations with \( 10^6 \) time steps.

It must be remarked that the assumptions we make concerning mating choice and conditions for a newborn to
survive were adopted in order to capture some features of field observations and laboratory experiments of species which seem to speciate via disruptive selection on habitat/food preferences and assortative mating.

For instance, in a series of papers Rice and Salt \cite{15} presented experimental evidence for the possibility of sympatric speciation in *Drosophila melanogaster*. They started from the premise that whenever organisms sort themselves into the environment first and then mate locally, individuals with the same habitat preferences will necessarily mate assortatively. Others examples of sympatric speciation can be found for canids \cite{16}, lizards \cite{17} and pandas \cite{18}. In this latter example, the Giant Panda (*Ailuropoda melanoleuca*), and the Red Panda (*Ailurus fulgens*) are vegatarian carnivores that specialize in eating bamboo in Sichuan Province, China. The two species share the same habitats and bamboo plants. Both pandas feed on the same species of bamboo, but specialize in eating different parts of the bamboo plant. The Giant Panda feeds more frequently on bamboo stems, while the Red Panda feeds more frequently on bamboo leaves \cite{18}. In our simulation, disruptive selection explicitly arises from competition for a single resource (a potentially more common ecological situation). In this way, we may imagine, for instance, that the original population of *Drosophila melanogaster* is around 1/7 of this value.

To check the main responsible element for the observed speciation in our model, we have also performed several simulations considering all possible permutations of the four new aspects introduced into the usual aging model here studied:

1) the kind of species in the beginning of the simulation (only $N_0$ or $N_1$ or $N_2$);

2) Mating preferences: The mating occurs only among individuals of the same species or it occurs without taking into account the individual’s species (randomly);

3) Back mutation: A reversal process whereby a gene that has undergone mutation returns to its previous state, i.e., if the randomly chosen position for introducing mutation at birth is the bit identifier of the species (position 11), with a small probability $p_{bn}$, a bit set to 1 in the parent’s genome, it is set to 0 in the offspring genome. However, if this bit is already set to 0, it remains 0 in the offspring genome. For all the others positions in the genetic strand only the possibility of harmful mutation is considered.

4) With or without the coexistence rule: the newborn with $n = 0$ ($n = 2$) can occupy an empty site or a site already occupied by a single individual with $n = 2$ ($n = 0$),
or they can born if and only if there is an empty site to be placed.

Additionally, for all the cases, we also analyzed if there would be any finite size effects related to the number of sites which the newborn can be placed: the newborn can be randomly placed only on the nearest-neighbor mother’s site or in any site of the lattice.

If the simulation starts only with individuals \( N_0 \) or only with individuals \( N_1 \), and only harmful mutations are introduced, the speciation is observed only if the coexistence rule is considered. However, for the case starting with only individuals \( N_2 \), the time evolution of the population presents only individuals \( N_2 \), since if only deleterious mutations are considered one bit set to 1 does not change to one set to 0. That’s why in this case, the speciation could be observed only when the back mutations with \( p_{bm} = 5 \times 10^{-5} \) were allowed. Furthermore, in all cases starting only with species \( N_2 \), the newborns suffer back-mutations with a small probability \( p_{bm} \).

If there is no mating preference, which means that individuals from different species can mate with one another, then the speciation occurs, however the coexistence rule must be considered, as mentioned before. The results obtained when the newborns are allowed to occupy only the nearest-neighbor mother’s sites and those ones when it could also occupy any lattice site show qualitatively the same behavior. It has been also noticed that the vanishing of intermediate individuals \( (N_1) \) (Fig. 1) was not caused by the restriction that their offspring could be born only if there is an empty site in the nearest-neighbor mothers’ site, since simulations taking into account the nearest-neighbor, the next-nearest-neighbor mother’s site and any empty site of the lattice (independently of its neighborhood) did not avoid their extinction, which, in fact, was observed to be due to the competition with the others two species \( (N_0 \text{ and } N_2) \) for the same resources, that is imposed through the prohibition of newborns \( N_2 \) to born on sites which are already occupied. This conclusion could be reinforced with the findings of the following investigation: During a certain timestep \( t < t_1 \), the population evolves considering the possibility to put newborns on empty sites and also the coexistence rule. For \( t \geq t_1 \), when the population size has reached the equilibrium (constant in time), the coexistence rule is disregard and only one condition for the survival of the offspring is hold: the newborn will be born only if there exists any empty lattice site (not only in the mother’s neighborhood) for it to be placed on. As we can see from the Figure 3 when \( t < t_1 \), the sympatric speciation occurs and the results are similar to one showed in Figure 1 \( (N_0 \approx N_2) \gg N_1 \). However, for \( t \geq t_1 \), since the population size of the mixed species is larger than those with \( n = 2 \ (N_0 > N_1 > N_2) \), it is not possible anymore to affirm the existence of speciation, even the coexistence of two separate species in the same environment (site). Based on this result, we can conclude that the sympatric speciation obtained in our simulations is essentially caused by non-existence of competition between the newborns with \( N_0 \) and \( N_2 \) and not by the finite size effects related to the neighborhood considered to place the newborns \( N_1 \).

If the females mate with males from any species randomly, the coexistence rule is the only condition required for occurring speciation, excepting when the simulation starts with species \( N_2 \) we must also take into account the back-mutations, for the reason previously mentioned. From this result, we can conclude that the mating preference condition is not necessary to be imposed in order to obtain sympatric speciation in our model, but the species’s ecological behaviour introduced through the coexistence rule. This conclusion is more reinforced with the results from all the simulations discussed before in which the coexistence rule always must be considered to lead the population to speciation.

As a final study, we examine our populations mortalities. In 1825, based on observed death and population records of people in England, Sweden, and France between ages 20 and 60 in the nineteenth century, the British actuary Benjamin Gompertz derived a simple formula describing the exponential increase in death rates between sexual maturity and extreme old ages [20]. This formula, \( q(\text{age}) = A \times \exp(b \times \text{age}) \), is commonly referred to as the Gompertz’s law of mortality. As Fig. 2 shows, our results for the mortality above the minimum reproduction age \( R = 8 \), are in a good agreement with the Gompertz law.

4 Conclusions

In conclusion, the results presented here are based on a very simple assumption that a single locus in the individual’s genome determines the ecological behavior of the
individual and identifies its species, in the framework of the Penna bit-string model. Despite its simplicity and limited applicability, our results clearly show the emergence of sympatric speciation in diploid sexual age-structured populations of individuals that are distributed on a square lattice. The introduction into the model of a specific gene responsible for the ecological behaviour of the individual: individuals with \( n = 1 \) and \( n = 2 \) do not compete for food resources, so they can bunch together at the same lattice site, tailored to reproduce recent observations of existing species \(^{21}\) that have led to the suggestion that ecological adaptation is the driving force behind divergence of populations leading to speciation: a gene, \( \text{desaturase}_2 \), of \( \textit{Drosophila melanogaster} \), which confers resistance to cold as well as susceptibility to starvation, underlies a pheromonal difference and contributes to the reproductive isolation between some \( \textit{Drosophila} \) species: the Zimbabwe and Cosmopolitan races.

Since the speciation discussed in this paper is triggered by interaction or competition between organisms, and not merely by mutation, the process is not so much random as deterministic. In fact, the speciation process occurs irrespective of the adopted random number in the simulation, as well without assuming mating preference. According to our scenario, the coexistence of the two groups \( (N_0 \text{ and } N_2) \) is the main mechanism that often leads the population to speciation. It should be also noted that the change in genotypes occur within few generations. The speed of genetic change, of course, depends on the mutation rate, but the present mechanism is found to work even for any smaller mutation rate (say \( T \leq 5 \)).

The relationship between our model and previous physics models, like directed percolation or reaction-diffusion problems, as suggested by an anonymous referee, unfortunately remains to be elucidated, even though some relations to reaction-diffusion systems can be established, \textit{a priori}. By the way, further research is needed to explore that question, which it is not the goal and the context of the present paper.

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