

Adaptive peak shifts with stress-induced mutation

Yoav Ram and Lilach Hadany

23 April 2013

1 Introduction

In 1931, Sewall Wright presented a problem which is still an open problem in evolutionary biology (Wright 1931). Given that the selective value of many alleles depends on other alleles, Wright suggested that "Two superior combinations that differ by two or more gene replacements may both be superior to the intermediate ones" (Wright 1988). The solution Wright suggested was called the "shifting-balance theory" [REF] and is still taught today in many "Introduction to Evolution" courses. This solutions appears to be valid (Crow et al., 1990; Wade and Goodnight, 1991; Coyne et al., 1997) but it seems that the range of parameters for which it works is limited (Moore and Tonsor, 1994; Gavrillets, 1996; Phillips, 1996).

Mutation is a major factor in this process: It creates the new alleles which later fix. If creating new favorable alleles was the only effect of mutation on evolution, a high mutation rate would have been very favored, but of course most mutations are deleterious and the mutation rate is reduced by natural selection to very low levels (Kimura 1967; Lynch 2011).

However, stress-induced mutation, in which stressed individuals increase their mutation rates, is an exception to this rule. In a previous work we have shown that stress-induced mutation is likely to evolve due to natural selection in asexual populations and that it increases the mean fitness of populations due to the increased generation of beneficial mutations in unfit individuals. Additionally, stress-induced mutation has been demonstrated in various species, both prokaryote and eukaryote (Galhardo, Hastings, and Rosenberg 2007; Sharp and Agrawal 2012; MacLean, Torres-Barceló, and Moxon 2013).

Here, we analyze a simple population genetic model of an asexual population with two bi-allelic loci. We derive analytical expressions that suggest that stress-induced mutation greatly increases the population adaptation rate. We use stochastic simulations to validate our analytic approximations.

2 Analytical model

Consider the two bi-allelic loci A/a and B/b and a population that reached a mutation-selection balance (MSB) in an environment in which ab is the optimal genotype with a fitness value of 1, single mutants (Ab and aB) suffer

from a selective disadvantage s and have a fitness value of $1-s$, and double mutants (AB) have a fitness value of $(1-s)^2$. This corresponds to a fitness function in which the effect of deleterious mutations are independent of each other and therefore the fitness of an individual is $(1-s)^m$ where m is the number of deleterious mutations the individual has accumulated.

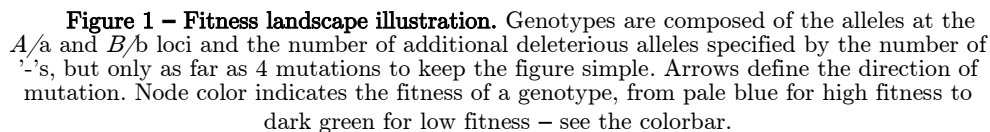
Mutation from a to A and from b to B occurs with a probability μ at reproduction and we disregard back-mutation. In addition, new deleterious mutations occur across the genome at reproduction, and the number of such mutations follows a Poisson distribution with a mean U . Although there is a direct relation between U and μ (for example, $\mu=U/5000$), having two separate parameters helps to distinguish between the two effects of mutation on adaptive evolution – the generation of beneficial mutations (μ) and the generation of deleterious mutations (U).

We define stress-induced mutation as the case in which an individual with a fitness below 1 hypermutates, increasing both his mutation rates τ -fold.

At the MSB, the frequency of wildtype (ab) individuals is $1-\mu/s+O(\mu^2)$, the frequency of single mutants (Ab and aB combined) is $2\mu/s+O(\mu^2)$ and the frequency of double mutants (AB) is $(\mu/s)^2+O(\mu^3)$ [REF].

To incorporate random sampling into the model, we denote the population size by N .

We are interested in the ability of the population to adapt to a new environment in which the genotype AB inhabits a new adaptive peak with a fitness value of $1+sH$, where H scales the height of the new peak in comparison with the existing peak at ab . See Figure 1 for an illustration of the two-peak fitness landscape.



There are several considerations on the relations between the main parameters:

- Summarizing the above requirements for stress-induced mutation to have an effect on adaptation in this model we get:

$$U < s \quad (1)$$

$$\frac{s}{\mu} < N < \left(\frac{s}{\mu}\right)^2 \quad (2)$$

For the bacteria species *Escherichia coli* estimations of the selection coefficient and mutation rates are $s=0.01$ (Kibota and Lynch 1996), $U=0.0004$ (Wielgoss et al. 2011) and $\mu = 8 \cdot 10^{-8}$ (Gordo, Perfeito, and Sousa 2011) which yield a fairly reasonable constraint on the population size - $1.25 \cdot 10^5 \leq N \leq 1.5 \cdot 10^6$.

2.2 Appearance of a double mutant

Assuming **Error! Reference source not found.**, there are no double mutants (AB) at the time of the environmental change. New double mutants can appear either via a double mutation in a wildtype individual (ab) or via a single mutation in a single mutant (Ab or aB). At the MSB the number of deleterious mutations follows a Poisson distribution (Haigh 1978). Therefore, the frequencies of mutation-free wildtype and single mutants are roughly $e^{-\frac{U}{s}}$ and $\frac{\mu}{s} e^{-\frac{\mu}{s}}$. The fitness of wildtype and single mutants is 1 and $1-s$, respectively. We assume that individuals with mutations other than in the A/a and B/b loci are "evolutionary dead ends". Assuming that mutation is a Poisson process, only a fraction e^{-U} of the individuals are of interest. All together, the probability that a newborn is a double mutant can be summarized by:

$$q = \mu^2 e^{-\frac{U}{s}-U} + 2 \frac{\mu^2}{s} e^{-\frac{\mu}{s}-U} \quad (3)$$

However, if mutation is stress-induced, then the mutation rate of single mutants is increased τ -fold and the above probability is:

$$q_{sim} = \mu^2 e^{-\frac{U}{s}-U} + 2 \frac{\tau \mu^2}{s} e^{-\frac{\mu}{s}-\tau U} \quad (4)$$

Note that stress-induction only increases the transition from single mutants to other types, but does not change the MSB frequency of single mutants, because that is determined by the mutation rate of the wildtype.

If the mutation rate of wildtype is increased as well, as in the case of constitutive hypermutation, the probability changes to

$$q_{cm} = (\tau \mu)^2 e^{-\frac{\tau U}{s}-\tau U} + 2 \frac{(\tau \mu)^2}{s} e^{-\frac{\tau \mu}{s}-\tau U} \quad (5)$$

because increased mutation rate in the wildtype affects the MSB frequencies.

2.3 Fixation of a double mutant

Assuming that the advantage of the double mutant is considerable (for example, $H > 1$) it has two possible fates after its appearance: fixation and extinction. Following Eshel (1981) the fixation probability of the double mutant is

$$\pi = 2 \frac{\alpha - 1}{\alpha} + o(\alpha - 1) \quad (6)$$

where α is the relative fitness of the double mutant normalized by the population mean fitness:

$$\alpha = \frac{(1 + sH)e^{-U}}{\bar{\omega}} \quad (7)$$

and assuming that fitness is measured by the number of progeny which is Poisson distributed.

Because the frequency of double mutants is very low at the stage where they are subject to possible extinction by drift, the population mean fitness can be calculated without considering them, so the value we use is the mean fitness of the population at the MSB. Without stress-induced mutation, this evaluates to $\bar{\omega} = e^{-U}$ (Kimura and Maruyama 1966). Therefore, $\alpha = 1 + sH$ and

$$\pi = \frac{2sH}{1 + sH} + o(sH). \quad (8)$$

Assuming that $1 - \alpha = sH$ is small we get

$$\pi^* \approx 2sH \quad (9)$$

Which is a classic result in population genetics [REF]. Note that this value does not depend on the mutation rate, thus it is not affected by constitutive mutation.

However, as we have shown before (Ram and Hadany 2012), the mean fitness of a population with stress-induced mutation can be different this value if beneficial mutations are allowed. Here, the mean fitness with stress-induced can be calculated by separating the population to the wildtype fraction $e^{-U/s}$ which has fitness 1 and the non-wildtype fraction $1 - e^{-U/s}$. Within the non-wildtype subpopulation, which have at least one mutation, additional mutations are Poisson distributed with expectation $\tau U/s$ because this subpopulation is hypermutating. Therefore the mean fitness of this subpopulation is $(1 - s)e^{-\tau U}$. Taken together, the mean fitness of a population with stress-induced mutation is

$$\bar{\omega}_{sim} = e^{-\frac{U}{s}} + (1 - e^{-\frac{U}{s}})(1 - s)e^{-\frac{\tau U}{s}} \quad (10)$$

Plugging this in (6) and (7) gives a different fixation probability for populations with stress-induced mutation (see Appendix 6.1 for full derivation):

$$\pi_{sim} = 2 - 2 \frac{e^{U - \frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{U - \tau U}}{1 + sH} \quad (11)$$

Interestingly, the above can indeed be larger than $2sH$. This is because the relative fitness of the double mutant in a population with stress-induced mutation is greater than without – single mutants and other individuals below both adaptive peaks (ab and AB) will hypermutate and increase their mutational load.

2.4 Total adaptation time

From the probability that in a population without double mutants a newborn is a double mutant we can derive the probability that a double mutant would appear in the population: $1 - (1 - q)^N$. The condition (2) guarantees that Nq is very small, hence this probability can be approximated by Nq using the Binomial series expansion.

Once a double mutant appears it has a probability π to go to fixation.

The time for adaptation T_a can roughly be approximated by the waiting time for a double mutant which will go to fixation T_w . This is true as long as fixation is a much faster process than mutation ($\mu^2 < s$). T_w has a geometric distribution with parameter $Nq\pi$ and therefore the expected time for adaptation can be approximated by

$$E[T_a] \approx 1/Nq\pi \quad (12)$$

which can be evaluated with or without stress-induced mutation by using (3), (4), (9) and (11).

3 Stochastic simulation

Our analytical model produces approximations for the adaptation time with and without stress-induced mutation. To facilitate these approximations several assumptions were made:

1. Mutation must be weaker than selection – Eq. (1)
2. The population size must be within a specific range, although this range may be quite large – Eq. (2)
3. Individuals with deleterious mutation, except at the A/a and B/b loci, are "Evolutionary dead ends".

To verify that the our approximations are correct, we developed a Wright-Fisher model with mutation, selection and random drift. The main differences between the analytical model and the stochastic simulations are described here, for more details on the design of the simulations please refer to the Methods section.

There are several advantages to stochastic simulations: (i) The simulations incorporate genetic drift by random sampling each generation from the previous one using a multinomial distribution. (ii) Individuals with deleterious mutations are not "evolutionary dead ends" – individuals are allowed to accumulate up to $G=25$ deleterious mutations. (iii) Simulations start with a wildtype, mutation-free population on a single-peak adaptive landscape and after reaching MSB a new adaptive peak is introduced for the AB genotype. Therefore, the stochastic model assumes nothing about the distribution of deleterious mutations at the MSB. (iv) We ran simulations in which selection and mutation were at the same order of magnitude, as well as simulations in which the population size is lower or higher than the constraints require (Eq. (2)).

Of course, an analytical model is preferable to simulations as it allows us to reach general conclusions where simulations only allow us to statistically estimate results for specific parameter sets. In addition, the simulations are computationally demanding, each running for up to several hours.

4 Results

4.1 Analytical approximations

4.2 Appearance of a double mutant

Our analytical approximation from the Analytical model section can be further simplified by using first-order approximations (see Appendix 6.2).

For the probability that a newborn individual is a double mutant we get the following approximations:

$$q^* = (1 - U)(2 - U)\frac{\mu^2}{s} \quad (13)$$

$$q_{sim}^* = (1 - \tau U)2\frac{\tau\mu^2}{s} \quad (14)$$

It is interesting to see that the first term is larger without stress-induction but the middle and last terms are both larger with stress-induction. Figure X shows that for a wide parameter range, q_{sim}^* is larger than q^* .

4.3 Fixation of a double mutation

The first-order approximation of the fixation probability of a double mutant with and without stress-induced mutation are (see Eq. (9) and Appendix 6.2):

$$\begin{aligned} \pi^* &= 2sH \\ \pi_{sim}^* &= \pi^* + 2\frac{U^2}{s}(1 - s)(\tau - 1) \end{aligned} \quad (15)$$

Because $s < 1$ and $\tau > 1$ these approximations suggest that $\pi_{sim} > \pi$, that is, stress-induced mutation increases the fixation probability of the double mutant. This is because stress-induced mutation increases the genetic variation in the population due to higher mutation rates in maladapted individuals. This effect increases with the mutation rate U and with the mutation rate increase τ but decreases with selection s . Note that this effect does not depend on beneficial mutation (μ).

4.4 Total adaptation time

The analytical approximation for the expected total adaptation time (Eq. (12)) is:

$$E[T_a] \approx 1/Nq\pi$$

The adaptation rate is the inverse of the expected adaptation time, Without stress-induced mutation we plug in Eqs. (9) and (13) to get:

$$E[T_a^*]^{-1} = Nq^*\pi^* = 2NH\mu^2(1-U)(2-U) \quad (16)$$

With stress-induced mutation we plug in Eqs. (14) and (15) to get:

$$E[T_{a_{sim}}^*]^{-1} = Nq_{sim}^*\pi_{sim}^* = 4N\tau\mu^2(1-\tau U) \left(H + \frac{U^2}{s^2}(1-s)(\tau-1) \right) \quad (17)$$

Comparing these two expression (see Appendix 6.3), the adaptation rate with stress-induced mutation is faster than it is without if

$$1 < \tau < 1/U \quad (18)$$

For example, for bacteria U is estimated to be between 0.0004 (Wielgoss et al. 2011) and 0.003 (Drake 1991), which sets the limit on τ to somewhere between 333-2,500. When τ is close to this limit, the mutation rate in hypermutating individuals is close to 1, which is probably much too high for single cell organism.

5 Methods

5.1 Simulations

We implemented a Wright-Fisher model using a simulation software written in Python (<http://www.python.org>), NumPy (<http://www.numpy.org>) and SciPy (<http://www.scipy.org>). The source code for the simulation is available on GitHub (<https://github.com/yoavram/ruggedsim>).

6 Appendix

6.1 Fixation probability with stress-induced mutation

$$\begin{aligned}
\pi &= 2 \frac{\alpha - 1}{\alpha} + o(\alpha - 1) \\
\alpha &= \frac{(1 + sH)e^{-U}}{\bar{\omega}} \\
\bar{\omega}_{sim} &= e^{-\frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{-\tau U} \\
\alpha_{sim} &= \frac{(1 + sH)e^{-U}}{e^{-\frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{-\tau U}} = \\
&\quad \frac{1 + sH}{e^U e^{-\frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{U-\tau U}} = \\
&\quad \frac{1 + sH}{e^{U-\frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{U-\tau U}} \Rightarrow \\
\pi_{sim}/2 &\approx \frac{\frac{1 + sH}{e^{U-\frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{U-\tau U}} - 1}{\frac{1 + sH}{e^{U-\frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{U-\tau U}}} = \\
&\quad \frac{1 + sH - e^{U-\frac{U}{s}} - \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{U-\tau U}}{\frac{1 + sH}{e^{U-\frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{U-\tau U}}} = \\
&\quad \frac{1 + sH - e^{U-\frac{U}{s}} - \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{U-\tau U}}{1 + sH} \Rightarrow \\
\pi_{sim} &\approx 2 - 2 \frac{e^{U-\frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1 - s)e^{U-\tau U}}{1 + sH}
\end{aligned}$$

6.2 First-order approximations

Here we derive the first-order approximations from the full analytical approximations described in the Analytical model section. The term "first-order" approximations is used here to describe the approximation of analytical expressions by linear expression or polynomials of the first degree. For example,

the expression e^{-U} can be written as a Taylor series $\sum_{i=0}^{\infty} (-1)^i \frac{U^i}{i!} = 1 - U + O(U^2)$. Therefore, when U is very small – for bacteria it is estimated to be between 10^{-4} and 10^{-2} – the linear expression $1-U$ is a good approximation for e^{-U} .

We will denote these first-order approximations by an asterix (*) added to the parameter symbol.

6.2.1 Appearance of a double mutant

Starting with Eq. (3) for populations without stress-induced mutation:

$$\begin{aligned} q &= \mu^2 e^{-\frac{U}{s}-U} + \frac{\mu^2}{s} e^{-\frac{\mu}{s}-U} = \mu^2 e^{-U} \left(e^{-\frac{U}{s}} + \frac{2}{s} e^{-\frac{\mu}{s}} \right) \approx \\ &\mu^2 (1-U) \left(1 - \frac{U}{s} + \frac{2}{s} \left(1 - \frac{\mu}{s} \right) \right) = \\ &(1-U) \left(\frac{\mu}{s} \right)^2 (s^2 - sU + 2s - 2\mu) \approx \\ &(1-U) \left(\frac{\mu}{s} \right)^2 (2s - sU) \end{aligned}$$

The last step assumes that $2s$ is much larger than s^2 and sU is much larger than 2μ . Rearranging the last expression gives us

$$q^* = (1-U)(2-U) \frac{\mu^2}{s}$$

For a population with stress-induced mutation the first-order approximation is based on the full expression in Eq. (4):

$$\begin{aligned} q_{sim} &= \mu^2 e^{-\frac{U}{s}-U} + 2 \frac{\tau \mu^2}{s} e^{-\frac{\mu}{s}-\tau U} = \mu^2 \left(e^{-\frac{U}{s}-U} + \frac{2\tau}{s} e^{-\frac{\mu}{s}-\tau U} \right) \approx \\ &\mu^2 \left(1 - \frac{U}{s} - U + \frac{2\tau}{s} \left(1 - \frac{\mu}{s} - \tau U \right) \right) = \\ &\frac{\mu^2}{s} \left(s - U - Us + 2\tau \left(1 - \frac{\mu}{s} - \tau U \right) \right) \approx \\ &\frac{\mu^2}{s} (s - U + 2\tau(1 - \tau U)) \end{aligned}$$

The last approximation assumes that Us is smaller than U and that τU is much larger than μ/s .

$$= \frac{\mu^2}{s} (s + 2\tau - U(1 + 2\tau^2)) \approx \frac{\mu^2}{s} (2\tau - 2\tau^2 U)$$

The last approximation assumed that $2\tau > s$ and $2\tau^2 > 1$, because $\tau > 1$ and probably even $\tau \geq 10$. Rearranging the last expression gives us the first order approximation for populations with stress-induced mutation:

$$q_{sim}^* = (1 - \tau U) 2 \frac{\tau \mu^2}{s}$$

6.2.2 Fixation of a double mutant

The fixation probability without stress-induced mutation is described and approximated in Eqs. (8) and (9).

With stress-induced mutation, the fixation probability (Eq. (11)) can be approximated by:

$$\begin{aligned} \pi_{sim}/2 &= 1 - \frac{e^{U-\frac{U}{s}} + \left(1 - e^{-\frac{U}{s}}\right)(1-s)e^{U-\tau U}}{1+sH} \approx \\ &= 1 - \frac{1+U-\frac{U}{s} + \left(1 - \left(1 - \frac{U}{s}\right)\right)(1-s)(1+U-\tau U)}{1+sH} = \\ &= 1 - \frac{1+U-\frac{U}{s} + \frac{U}{s}(1-s)(1+U-\tau U)}{1+sH} = \\ &= 1 - \frac{1+U-\frac{U}{s} + \left(\frac{U}{s} - U\right)(1+U-\tau U)}{1+sH} = \\ &= 1 - \frac{1+U-\frac{U}{s} + \frac{U}{s} - U + \frac{U^2}{s} - U^2 - \frac{\tau U^2}{s} + \tau U^2}{1+sH} = \\ &= 1+sH - 1 - \frac{U^2}{s}(1-s-\tau+\tau s) = \\ &= sH - \frac{U^2}{s}(1-s)(1-\tau) \Rightarrow \\ \pi_{sim}^* &\approx 2sH + 2\frac{U^2}{s}(1-s)(\tau-1) \end{aligned}$$

6.3 Comparison of adaptation rate

From Eqs. (16) and (17) we can derive the adaptation rate with stress-induced mutation in term of the rate without stress-induced mutation:

$$E[T_{a_{sim}}^*]^{-1} = 2\tau \frac{(1-\tau U)}{1-U} E[T_a^*]^{-1} + 4N\tau\mu^2(1-\tau U) \frac{U^2}{s^2}(1-s)(\tau-1)$$

Now if $2\tau \frac{(1-\tau U)}{1-U} > 1$ and because the second term is positive then we can infer that the rate with stress-induced mutation is faster than without. This condition can be rewritten:

$$2\tau(1-\tau U) > 1-U \Rightarrow 2\tau^2 U - 2\tau + 1 - U < 0$$

Using the quadrate formula this translates to:

$$2\tau U < \sqrt{\frac{2U^2 - 2U + 1}{U^2}} U + 1 \Rightarrow 2\tau U < \sqrt{2 + \frac{1-2U}{U^2}} U + 1$$

Because U is very small, $1-2U$ is well approximated by 1 , and $1/U^2$ is much larger than 2 , so this can be approximated by

$$2\tau U < \sqrt{\frac{1}{U^2}}U + 1 = 2 \Rightarrow \tau U < 1 \Rightarrow$$

$$\tau < \frac{1}{U}$$