

Population genomics of rapid adaptation by soft selective sweeps

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Organisms can often adapt surprisingly quickly to evolutionary challenges, such as the application of pesticides or antibiotics, suggesting an abundant supply of adaptive genetic variation. In these situations, adaptation should commonly produce ‘soft’ selective sweeps, where multiple adaptive alleles sweep through the population at the same time, either because the alleles were already present as standing genetic variation or arose independently by recurrent *de novo* mutations. Most well-known examples of rapid molecular adaptation indeed show signatures of such soft selective sweeps. Here, we review the current understanding of the mechanisms that produce soft sweeps and the approaches used for their identification in population genomic data. We argue that soft sweeps might be the dominant mode of adaptation in many species.

Hard and soft selective sweeps

Rapid adaptation has long been associated primarily with situations where selection is acting on quantitative traits that are highly polygenic, for example during breeding experiments. Such traits can respond quickly to changing selective pressures via small adjustments in the population frequencies of a large number of already present polymorphisms [1]. Under this so-called ‘infinitesimal model’ [2], adaptation is expected to leave subtle signatures in population genomic data because the underlying polymorphisms may have existed long enough in the population to become unlinked from their surrounding genetic variation.

However, recent studies show that rapid adaptation can often involve only few alleles of large individual effect that were previously rare or even absent in the population. Prominent examples include the evolution of pesticide resistance in insects [3], color patterns in beach mice [4], freshwater adaptation in sticklebacks [5], and lactose persistence in humans [6]. Our standard model for describing the population genetics of adaptation in these cases is the so-called ‘selective sweep’ [7,8]. In contrast to the infinitesimal model, in a selective sweep the adaptive alleles were previously rare, are still in linkage disequilibrium (LD)

with surrounding genetic variation, and change their population frequencies substantially due to positive selection.

Selective sweeps can be ‘hard’, where a single adaptive allele sweeps through the population, or ‘soft’, where multiple adaptive alleles at the same locus sweep through the population at the same time [9]. By definition, whether a sweep is hard or soft in a given population sample is determined by the genealogy of adaptive alleles at the selected site: in a hard sweep, the lineages in the sample that carry the adaptive allele coalesce more recently than the onset of positive selection, that is, the point in time when it first became advantageous to carry the allele (Figure 1A). By contrast, in a soft sweep, they coalesce before the onset of positive selection. Thus, sweeps in which several adaptive mutations of independent origin are present in a sample should be soft in most cases, regardless of whether the mutations arose *de novo* after the onset of positive selection (Figure 1B) or were already present previously as standing genetic variation (Figure 1C, top row). However, a situation where the adaptive allele arose only once but reached some frequency before the onset of positive selection, and several copies then swept through the population, is still considered a soft sweep if the lineages coalesce before the onset of positive selection (Figure 1C, bottom row).

Note that the above definition of hard and soft sweeps is based on a population sample. Thus, it is possible that the same adaptive event can yield a soft sweep in one sample but remain hard in another, depending on which alleles were sampled. For instance, if one had chosen only the six blue individuals from Figure 1B, the sweep in this subsample would have been hard. We also want to emphasize that the notion of a selective sweep that we adopt in this review refers solely to the population dynamics of adaptive alleles at the particular locus and the resulting signatures in population genomic data. This definition does not consider the actual molecular nature of the involved mutations, which may often be unknown. One consequence of this definition is that, in principle, the different adaptive mutations that contribute to a sweep at a given locus do not have to result from the same selective pressures. We explain this in more detail below when we discuss the definition of the relevant genetic locus for a selective sweep.

Whether adaptation produces hard or soft sweeps depends primarily on the availability of adaptive mutations [9,10]. Hard sweeps are expected when adaptive alleles are not present in the population at the onset of selective pressure and when the waiting time for adaptive

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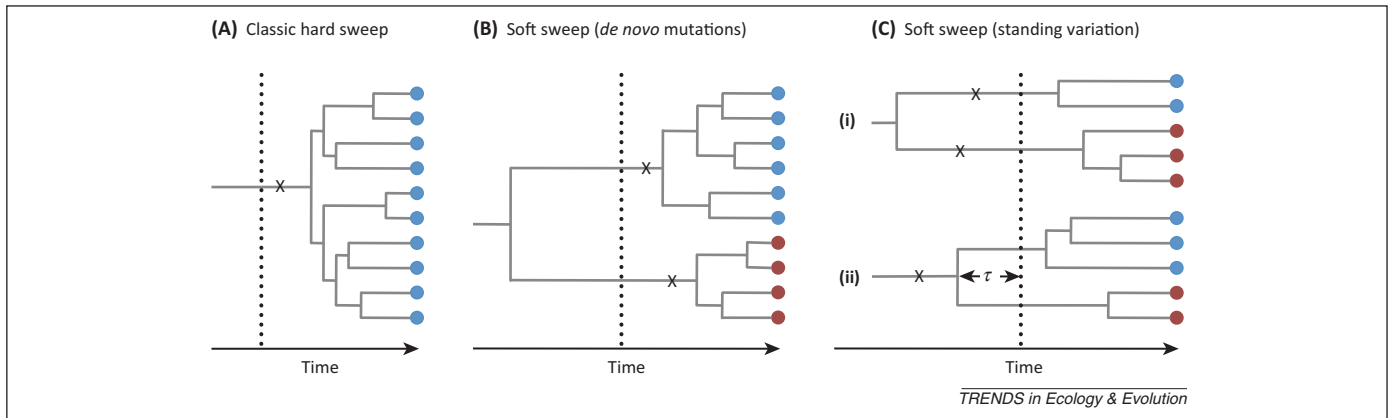


Figure 1. Definition of hard and soft sweeps. **(A)** In a hard sweep, all adaptive alleles in the sample arise from a single mutation (depicted by x) and coalesce after the onset of positive selection (broken line). Note that, even if the mutation had arisen before the onset of positive selection and was present as standing genetic variation, this would still be considered a hard sweep as long as only a single lineage is ultimately present in the sample. **(B)** In a soft sweep from recurrent *de novo* mutations, the adaptive alleles in the sample arose from at least two independent mutation events after the onset of positive selection and the lineages coalesce before the onset of positive selection. **(C)** In a soft sweep from the standing genetic variation, adaptive alleles were already present at the onset of positive selection. The different lineages in a population sample can originate from independent mutation events (i) or from a single mutation that reached some frequency before the onset of positive selection, such that several copies present at that time then swept through the population (ii). In this latter case, the population genetic signatures of the sweep will depend on the time τ between coalescence and onset of positive selection. If τ is short, the sweep will appear similar to a hard sweep, whereas when τ is large, it will be similar to a soft sweep from several *de novo* mutations.

mutations is long. By contrast, soft sweeps are expected when the waiting time until an adaptive mutation arises is shorter than the time it takes for this mutation to spread through the population. This is the case: (i) in large populations; (ii) when adaptation has a large mutational target (for example, when every loss-of-function mutation in a gene is adaptive) [11]; or (iii) when adaptation utilizes alleles present as standing genetic variation, either in mutation–selection–drift balance or maintained by balancing selection [12,13]. Soft sweeps are also possible as a result of parallel adaptation in geographically structured populations when several mutations emerge independently in distant locations before one has spread over the entire range [14–17]. In this case, ‘local’ samples from a subpopulation might always yield hard sweeps, whereas ‘global’ samples across subpopulations can yield soft sweeps.

In a hard selective sweep, all lineages collapse into a single cluster, generating characteristic signatures in population genomic data, such as a reduction in genetic diversity around the adaptive site [7,8,18], an excess of high-frequency derived alleles and singletons [19–22], and the presence of a single, long haplotype [23]. These hallmark signatures underlie most commonly used approaches for identifying sweeps [18,19,24–31]. By contrast, in a soft sweep, lineages collapse into more than one cluster and several haplotypes can be frequent in the population at the adaptive locus. Thus, diversity is not necessarily reduced and deviations in the frequency distributions of neighboring neutral polymorphisms are typically weak compared with hard sweeps [13,32–34]. As a result, it is difficult to identify soft sweeps from polymorphism summary statistics, such as Tajima’s D [25], Fay and Wu’s H [19], and the composite likelihood ratio (CLR) test [35].

Scans for positive selection in population genomic data have typically focused on identifying hard sweeps and have only limited power for soft sweeps [13,32,33]. Hence, if soft sweeps are pervasive, then most of them should have evaded detection and we might be missing an entire class of important adaptive events.

Signatures and examples of soft sweeps

There is mounting evidence, both from individual case studies and genome-wide scans, that soft sweeps are indeed common in a broad range of organisms, from viruses to insects and even mammals. Below, we briefly review this evidence and discuss the diversity of approaches used to identify soft sweeps in molecular population genetic data.

Soft sweeps are abundant in case studies of adaptation

In some cases, it is possible to detect soft sweeps directly through the presence of adaptive mutations of independent origin. Figure 2A shows an example from the evolution of resistance to HIV treatment that involves a single amino acid change of the viral reverse transcriptase [36]. Penning et al. [37] analyzed viral samples obtained from the same patient before and after resistance had evolved. Before treatment, all viral samples were monomorphic for the lysine codon AAA at the resistance locus. After resistance had evolved, two different synonymous codons (AAT and AAC), both encoding asparagine, were frequent in the sample. This is a clear example of a soft sweep that could have originated either because both alleles were already present at the onset of treatment or from independent *de novo* mutations afterwards.

In a geographically structured population, sweeps that are hard in local samples can become soft in global samples that comprise individuals from geographically distant locations. This signature can then be used to infer cases of parallel adaptation [15,17]. The classic example for this scenario is lactase persistence in humans that evolved in parallel in Eurasia and Africa through independent mutations in the gene encoding lactase [6,38,39]. Figure 2B shows the length of homozygosity tracts flanking lactase persistence-associated single nucleotide polymorphisms (SNPs) in Eurasia and Africa from [38]. Within each region, lactase-persistent individuals show extensive haplotype homozygosity, sometimes extending over more than 2 Mb, whereas haplotype homozygosity in nonpersistent haplotypes is not elevated, suggesting hard sweeps in both

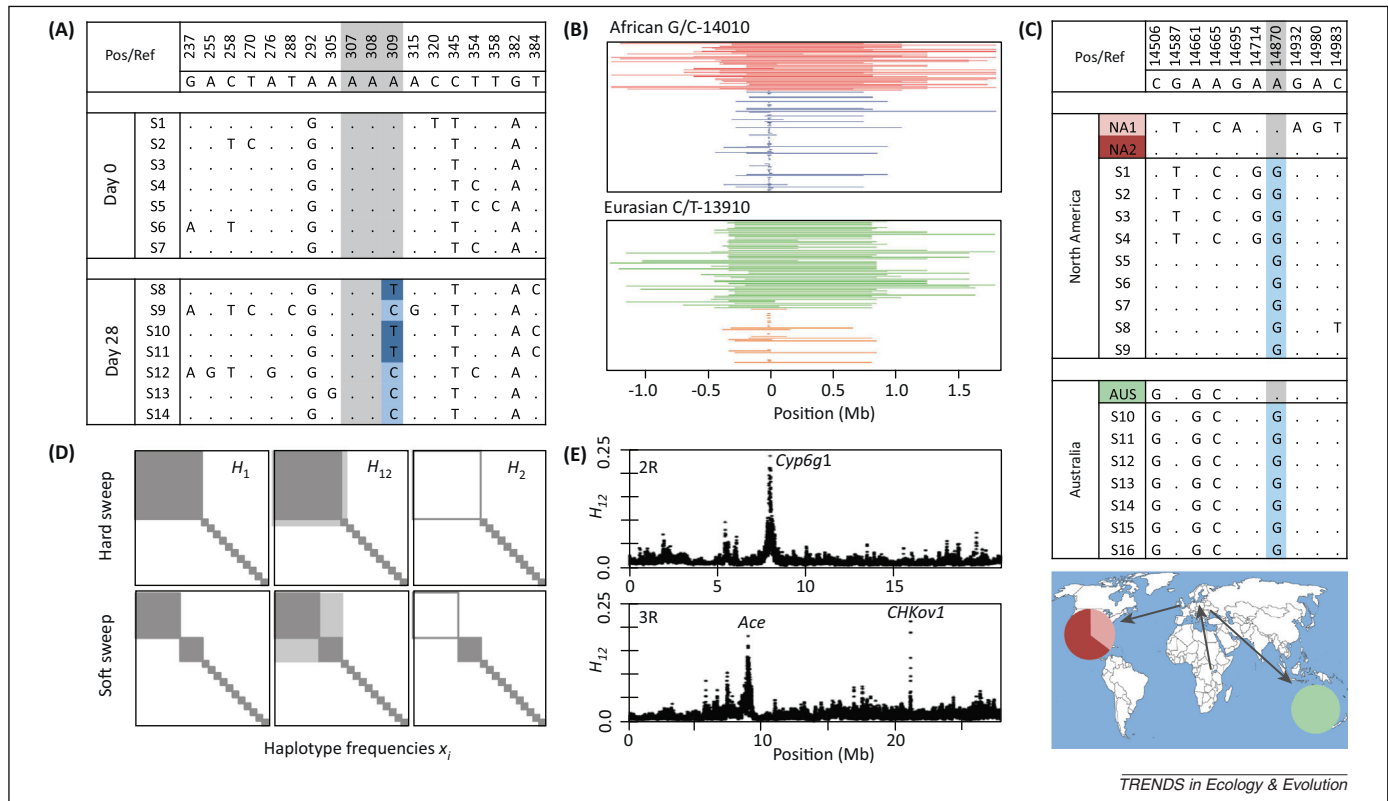


Figure 2. Soft sweep examples in population genomic data. **(A)** Haplotypes of the HIV reverse transcriptase observed in two samples taken from the same patient before treatment (day 0; samples S1–S7) and after resistance had evolved (day 28; samples S8–S14) from [37]. Treatment resistance involves a single amino acid change from lysine to asparagine in the codon spanning positions 307–309 (grey columns). The original AAA codon was replaced by a mixture of AAC and AAT codons that both encode asparagine. **(B)** Soft sweep in humans in the gene encoding lactase [38]. The top panel shows homozygosity tracts in African individuals that carry the persistent C-14010 allele (red) versus those that carry the nonpersistent G-14010 allele (blue). The bottom panel shows tracts for Eurasian individuals that carry the persistent T-13910 allele (green) versus those that carry the nonpersistent C-13910 allele (orange). **(C)** Soft sweep during the evolution of pesticide resistance in *Drosophila melanogaster* [41]. The table shows the observed haplotypes in a region of the *Ace* gene from flies sampled in North America and Australia. *D. melanogaster* evolved in Africa and then spread worldwide via Europe (lower panel). The A to G mutation at position 14 870 of *Ace* increases resistance to several commonly used pesticides. NA1 and NA2 are commonly observed sensitive haplotypes in North America and samples S1–S9 show the haplotypes of nine resistant flies collected in North America. AUS is a commonly observed sensitive haplotype in Australia and sequences S10–S16 show the haplotypes of seven resistant flies collected in Australia. In both locations, resistance seems to have evolved on the locally prevailing sensitive haplotypes. **(D)** Haplotype homozygosity statistics. The top row depicts a hard sweep with a single common haplotype and several low-frequency variants; the bottom row depicts a soft sweep with two common haplotypes. The total gray area in the left panel specifies haplotype homozygosity $H_1 = \sum_i x_i^2$. The middle panel shows extended haplotype homozygosity H_{12} , obtained after combining the frequencies of the two most common haplotypes. The right panel shows haplotype homozygosity calculated after removing the most frequent haplotype. H_1 is larger (and H_2 smaller) for the hard sweep than for the soft sweep. H_{12} is similar in both scenarios. **(E)** H_{12} scan for chromosomes 2R and 3R of *D. melanogaster* [64]. The three most prominent peaks coincide with three well-known cases of adaptation at the loci *Cyp6g1*, *Ace*, and *CHKov1*.

regions with some ancestral variation remaining. However, the persistent haplotypes in the two regions are highly divergent from each other, indicating independent origins of the adaptive mutations and a global soft sweep on the scale of the entire population [38].

Another clear example of parallel adaptation is the evolution of pesticide resistance in *Drosophila melanogaster* through mutations in the gene *Ace*, which encodes acetylcholinesterase, a major target of commonly used pesticides [40]. Karasov *et al.* [41] found that the same adaptive mutation, a mutation from A to G at position 14 870, resides on distinct haplotypes depending on the geographic locations from which the flies were sampled (Figure 2C). Specifically, in flies from North America, the resistance mutation is located on a haplotypic background that is also common among the sensitive flies in North America, whereas in flies from Australia, the resistance mutation resides on a background that is common among sensitive flies in Australia, but rare in North America. Thus, it appears that the resistance mutation arose independently on the haplotypic backgrounds that are common

on each continent. Moreover, multiple resistant haplotypes at *Ace* are also present within continents, including haplotypes with resistant mutations at two other sites within the gene [41]. The evolution of pesticide resistance at *Ace* provides a clear example of soft sweeps being associated with rapid, multistep adaptation under strong selection, given that organophosphate pesticides have only been used since the 1950s.

Soft sweeps have recently been observed in various other case studies of adaptation. For example, in malaria parasites, multiple *de novo* amplification events of the gene encoding *Plasmodium falciparum* multidrug resistance protein 1 (*pfmdr1*) confer resistance to mefloquine [42], whereas multiple independent mutations in the genes encoding dihydrofolate reductase (*dhfr*) and deoxyhypusine synthase (*dhps*) confer resistance to sulfadoxine pyrimethamine [43]. Several studies have observed soft sweeps during the evolution of drug resistance in HIV [44,45]. Fungicide resistance of the plant pathogen *Plasmopara viticola* arose via four independent *de novo* mutations in the gene encoding cytochrome *b* [46]. The evolution of resistance to

benzimidazole in *Teladorsagia circumcincta*, a parasitic nematode of sheep, displays signatures of soft sweeps due to multiple independent mutations in the gene encoding β -tubulin [47]. The three best-understood cases of recent adaptation in *D. melanogaster* all produced soft sweeps: viral and organophosphate resistance at the *CHKov1* locus evolved from standing variation [48,49], whereas pesticide resistance at the *Ace* locus (as discussed above) as well as DDT resistance at the *Cyp6g1* locus, evolved via multiple independent *de novo* mutations [3,41,50,51]. In *Drosophila santomea*, soft sweeps have been observed from multiple *de novo* loss-of-function mutations in the gene *tan* [52]. In the mosquito *Culex pipiens*, multiple independent duplications of the *ace-1* locus provide insecticide resistance [53]. The parallel evolution of the freshwater-specific reduction of armor plates in sticklebacks produced soft sweeps [5,54,55], as did adaptation at the *Mc1r* locus in mice [4,56]. A clear example of a soft sweep in the brown rat are the several different allelic variants of the gene encoding vitamin K epoxide reductase complex subunit 1 (*VKORC1*) that rapidly evolved in response to the rodenticide warfarin [57,58]. Additional prominent examples in humans are the different variants of the sickle cell allele in β -globin [59,60], the several mutations in the gene encoding glucose-6-phosphate dehydrogenase (*G6PD*) that evolved independently in response to malaria [61], and adaptation from standing genetic variation in the genes encoding Abnormal spindle-like microcephaly-associated protein (*ASPM*) and prostate stem cell antigen (*PSCA*) [62].

Soft sweeps are abundant in systematic genomewide scans for adaptation

Even though soft sweeps are expected to leave more subtle signatures in population genomic data compared with hard sweeps, it is still possible to distinguish them from patterns expected under neutral evolution. Depending on the degree of the ‘softness’ of the sweep, that is, the number of independently originated adaptive alleles in the sweep, only a few haplotypes may be frequent in a population sample. The two illustrations of soft sweeps in Figure 1, for instance, only have two components each (red and blue). Within each component, the coalescent resembles that of a hard sweep. As a result, LD is still higher than under neutrality [33,63] and, therefore, methods that detect perturbations in the haplotype structure, such as iHS [27,30] and XP-EHH [31], should retain some power to detect systematically soft sweeps in population genomic data as long as the sweeps were not too soft [33].

Messer and Neher [45] showed that it is possible to detect soft sweeps and distinguish them from hard sweeps using haplotype data. Their approach is based on the observation that, in a hard sweep and, thus, within each component of a soft sweep, the new variants of the adaptive haplotype that arise from mutation or recombination events during the sweep should be at low population frequencies (Box 1). In addition, these variants should typically differ from the original haplotype by only a single mutation or recombination event. However, two adaptive haplotypes from distinct components of a soft sweep can both be frequent and should also be more diverged from each other.

Box 1. Hard sweeps just looking soft?

How can one decide whether a sweep with several frequent haplotypes is truly a soft sweep, rather than just a hard sweep where recombination or mutation during the sweep has broken up the original haplotype into different variants? Assume that at time t_0 , an adaptive mutation establishes. Early during its sweep, mutation or recombination events on the sweeping haplotype can create new variants that also increase in frequency. Their expected frequencies, x_1, x_2, \dots , are determined by their seeding times, t_1, t_2, \dots , which gives rise to a characteristic frequency spectrum of haplotype variants in a hard sweep, as illustrated in Figure 1.

Messer and Neher [45] used branching process calculations to show that, for a hard sweep with selection coefficient s , the expected ratio between the frequency x_i of variant i and the frequency x_0 of the original haplotype is given by Equation 1:

$$E(x_i/x_0) \approx u/(is), \quad [1]$$

where u is the combined rate of mutation and recombination estimated over the whole locus. The approximation assumes that $u \ll s$. Thus, frequency distributions of haplotype variants in hard sweeps are described by a simple power-law: the most abundant adaptive haplotype, on average, is s/u times more frequent than the first variant, $2s/u$ times more frequent than the second variant, and so forth. For example, when recombination and mutation rate are both 10^{-9} per site per generation, then $u = 2 \times 10^{-4}$ for a locus of length 10 kb. In a hard sweep with $s = 0.01$, we then expect the original haplotype to be ≈ 50 times more frequent than its first variant.

Hence, even though the variance in x_1/x_0 can be large, the new variants in a hard sweep will, on average, be at low frequencies as long as loci are not too large [21,45]. By contrast, in a soft sweep, the frequencies of the most common and second most common haplotype can be similar and, therefore, x_1/x_0 much larger.

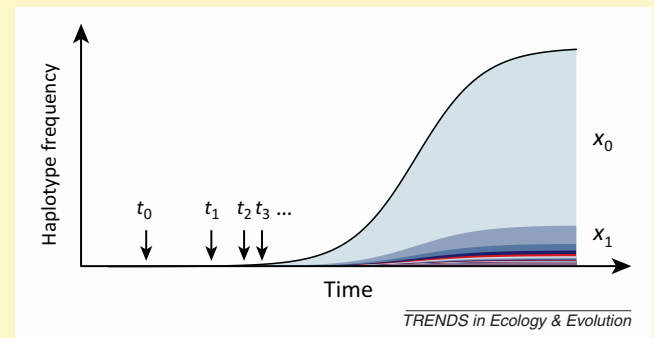


Figure 1. Haplotype frequency trajectories in a hard sweep.

Garud *et al.* [64] proposed a related approach to detect systematically both hard and soft sweeps in population genomic data and to distinguish them from each other. They developed a test statistic (H_{12}) that estimates haplotype homozygosity after combining the frequencies of the two most frequent haplotypes in a given genomic region (Figure 2D). Thus, a soft sweep with two frequent components is treated effectively as a hard sweep with one big component. H_{12} has high power to detect cases of recent and strong adaptation and, importantly, has similar sensitivity for both hard and soft sweeps, as long as the latter are not too soft and still comprise only a few frequent components. A genome-wide H_{12} scan in 192 sequenced *D. melanogaster* strains from North Carolina [65] revealed abundant signatures of recent adaptation with haplotype structure often extending over hundreds of kilobases (Figure 2E).

Garud *et al.* [64] also developed a second statistic (H_2/H_1) that compares haplotype homozygosities with

and without the most common haplotype (Figure 2D). High values of this statistic are expected only for soft sweeps. When they applied this test to the 50 most-prominent peaks in their H_{12} scan, all showed signatures of soft sweeps.

Likelihood of soft sweeps

The many examples reviewed above show that soft sweeps are common in a range of organisms. What are the circumstances under which this should be the case? In principle, soft sweeps can arise from recurrent *de novo* adaptive mutations, adaptation from the standing genetic variation, and parallel adaptation in geographically structured populations. Here, we discuss the key evolutionary parameters that determine the probabilities of soft and hard sweeps in each of these scenarios.

Recurrent *de novo* mutations

Consider a haploid population of size N in a Wright–Fisher model (i.e., constant population size, random mating, and discrete generations). Assume that a mutant allele confers a fitness advantage $s > 0$ over the wild type and arises at rate μ per individual per generation at the locus of interest. We define $\Theta = 2N\mu$ as twice the average number of new mutants that enter the population per generation. Given that only an approximate fraction $2s$ of new mutations will escape stochastic loss and successfully establish in the population [66], the rate of successfully establishing mutations is $2N\mu s \approx \Theta s$. Thus, the average waiting time for a successfully establishing mutation is $T_e \approx 1/(\Theta s)$. Once established, a mutant lineage grows approximately logistically, requiring $T_f \approx (1/s)\log(Ns)$ generations until fixation (assuming that there is no interference with other mutations) [7].

By comparing the timescales of establishment and fixation, we can estimate whether adaptation should primarily proceed from a single *de novo* adaptive mutation or involve multiple recurrent *de novo* mutations (Figure 3A): multiple origins should prevail whenever adaptive mutations enter the population frequently enough such that a second independent mutation can establish in the population before the first one has reached fixation and, thus, when $T_e < T_f$, yielding Equation 1:

$$\Theta > \frac{1}{\log(Ns)}. \quad [1]$$

This simple timescale argument already provides a key insight: whether adaptation is more likely to involve a single or multiple *de novo* mutations depends primarily on Θ , the rate at which adaptive mutations enter the population, whereas it depends only logarithmically on the strength of selection, because establishment and fixation time both scale inversely with s . Note that scenarios where $\Theta \geq 1$ should generally involve multiple *de novo* mutations unless positive selection is extremely weak.

Adaptation by multiple *de novo* mutations does not automatically produce soft sweeps in a finite population sample because one of the mutations might be at a particularly high population frequency and, thus, the only mutation present in the sample. Pennings and Hermisson [11] used coalescent theory to approximate the probability of

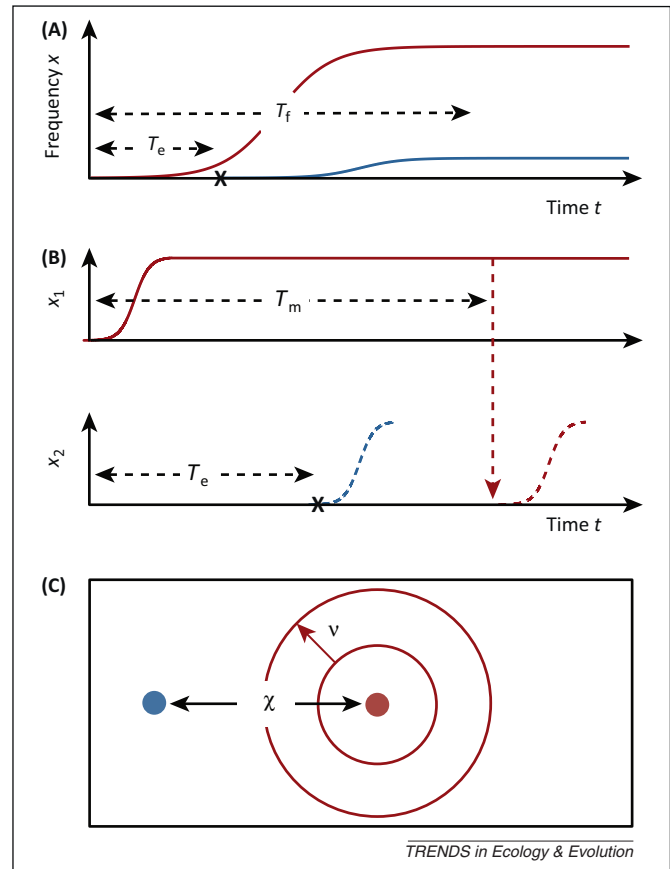


Figure 3. Likelihood of hard and soft sweeps and relevant timescales. **(A)** The red curve shows the frequency trajectory of an adaptive mutation. The blue curve shows the trajectory of another *de novo* adaptive mutation that successfully established before the first one became fixed in the population. This scenario is likely when establishment time T_e is shorter than fixation time T_f . **(B)** Adaptation in a subdivided population with two demes and migration. An adaptive mutation arises and sweeps through the first deme (red trajectory). The allele can subsequently migrate and also sweep in the second deme (broken red trajectory), resulting in a global hard sweep. Alternatively, an independent *de novo* adaptive mutation can arise first and sweep in the second deme (broken blue trajectory), resulting in a global soft sweep. **(C)** Adaptation in a spatially continuous population with limited dispersal. An adaptive mutation arises at one geographic location (red area) and then spreads through the population in a radial wave with speed v (red circles). While this mutation is still spreading, another *de novo* adaptive mutation arises at a different location that has not yet been covered by the first mutation (blue area). The characteristic length χ specifies the average distance traveled by an adaptive mutation until another successful mutation is expected to have arise within its already covered area.

observing multiple adaptive *de novo* mutations of independent origin in a random population sample of size n . To leading order, they obtained Equation 2:

$$P_{\text{soft}}(\Theta, n) \approx 1 - \prod_{i=1}^{n-1} \frac{i}{i + \Theta}. \quad [2]$$

Thus, even in small samples, adaptation should primarily lead to soft sweeps whenever $\Theta \geq 1$. For example, when $\Theta = 1$, we expect soft sweeps in samples of size $n = 10$ in 90% of the cases and, for $\Theta = 0.1$, we still expect soft sweeps in 25% of cases.

Adaptation from standing genetic variation

In sufficiently large populations, neutral and even deleterious mutations are present most of the time under mutation–selection–drift balance. When such mutations

suddenly become advantageous, adaptation can proceed from alleles that are already present as standing genetic variation. We can estimate the probability of a selective sweep from standing genetic variation using Equation 3:

$$P_{\text{sgv}} = \int_0^1 \rho(x)\Pi(x)dx, \quad [3]$$

where $\rho(x)$ is the probability density that the mutation previously segregated at frequency x in the population, and $\Pi(x)$ is the probability that a mutation with selection coefficient s , which is present at frequency x , eventually fixes in the population. For a previously neutral mutation under mutation–drift balance, this yields Equation 4 [9]:

$$P_{\text{sgv}} \approx 1 - \exp[-\Theta \log(2Ns)]. \quad [4]$$

Thus, a selective sweep from a preexisting neutral mutation is likely whenever $\Theta > 1/\log(2Ns)$, which is essentially the same condition we obtained for adaptation involving recurrent *de novo* mutations in Equation 1. Note that P_{sgv} is only marginally lower than the probability that the mutation is segregating in the population at all, $\Theta \log(N)$ for a neutral mutation [67].

For a mutation that was previously deleterious, P_{sgv} will always be smaller than for the neutral case because deleterious mutations, on average, segregate at lower frequencies and are present in the population less of the time than are neutral mutations. Specifically, if the mutation was previously deleterious with selection coefficient s' , one obtains Equation 5:

$$P_{\text{sgv}} \approx 1 - \exp[-\Theta \log(1 + R_\alpha)], \quad [5]$$

where $R_\alpha = 2Ns/(2Ns' + 1)$ is the relative selective advantage of the mutation [9]. In this case, adaptation from the standing genetic variation is likely only when $\Theta > 1/\log(1+R_\alpha)$.

The results from this and the previous section demonstrate that, in a panmictic population, selective sweeps involving previously neutral or deleterious standing genetic variation, as well as multiple *de novo* adaptive mutations, should be unlikely when $\Theta \ll 1$ and, therefore, most selective sweeps should be hard in this regime. However, when $\Theta \geq 1$, soft sweeps become common. In this regime, adaptive mutations are present in the population most of the time and the distinction between adaptation from standing genetic variation and recurrent *de novo* mutation becomes blurred.

It is also possible that adaptation commonly involves alleles from the standing genetic variation that are maintained by some form of balancing selection, for instance frequency-dependent selection, heterozygote advantage, or selection that varies systematically across time or space. If adaptation proceeds from such alleles, then its signatures will depend primarily on the number of initially present alleles at the locus that ultimately contribute to adaptation and their ages (Figure 1C): balanced alleles that have existed in the population for a long time will be present on diverse haplotypic backgrounds and may no longer be in LD with surrounding genetic diversity. When adaptation involves such alleles, it may only be visible in longitudinal data of polymorphism frequencies over time [68–71]. However, if the balanced alleles are still young and only present

on a few haplotypes, as has been proposed for adaptive walks in diploids [72], we can still expect to observe soft sweep signatures.

Parallel adaptation in structured populations

In a panmictic population, the speed at which an adaptive mutation spreads through the population is primarily determined by the strength of positive selection. However, in a spatially structured population, the spread of the adaptive mutation can be impeded by the time it takes until individuals with the mutation migrate into distant areas of the population. If this takes much longer than the establishment time of a new adaptive mutation in the population as a whole, then another adaptive mutation of independent origin can arise elsewhere in the population before the first adaptive mutation has covered the entire range. An example of such a scenario is the parallel evolution of lactase persistence in humans in Eurasia and Africa [6,38,39].

We can estimate the conditions under which parallel adaptation should be likely in structured populations by comparing the timescales of establishment and fixation. Here, the fixation time specifies the expected waiting time until an established adaptive mutation has covered the entire range of the population.

Let us first consider a simple scenario of a subdivided population with two panmictic demes consisting of N individuals each with migration rate m between them (Figure 3B). We further assume that $\Theta \ll 1$ within each deme, otherwise we already know that soft sweeps should be common even within demes and, thus, certainly in global samples comprising individuals across demes. In this regime, an adaptive mutation that establishes in the first deme will locally sweep to fixation in a shorter time than the waiting time $T_e = 1/(\Theta s)$, until an independent adaptive mutation is expected to establish in the second deme.

Migrants carrying the adaptive mutation from the first deme will establish in the second deme at an approximate rate $2s \times Nm$. The waiting time for this to happen is $T_m \approx 1/(2Nms)$. Parallel adaptation capable of producing soft sweeps in global samples should be likely when $T_m > T_e$ and, hence, when $\Theta > 2Nm$.

Thus, migration has to be extremely weak (i.e., lower than the mutation rate). Consider, for instance, a scenario with $\Theta = 0.01$, assuring that soft sweeps are not expected within demes. For parallel adaptation to become likely, migration would then need to be weaker than $2Nm = 0.01$. This means that, on average, only one individual would migrate between demes per 100 generations, corresponding essentially to two noninterbreeding populations.

The above model considered only two discrete demes. Ralph and Coop [16] investigated the other extreme case, that of a spatially continuous populations in which adaptive mutations arise locally and then spread through the population in a wavelike manner (Figure 3C). We can once again assess the conditions under which parallel adaptation becomes likely in this model from a simple comparison of timescales: Consider an idealized population with N individuals distributed evenly over a circular area of radius r . An adaptive mutation establishes at some location and then spreads in a radially expanding wave with constant speed v .

The average time until this mutation covers the entire area is on the order of $T_m \approx r/v$. Parallel adaptation should then be likely if again $T_m > T_e$, yielding Equation 6:

$$\Theta > \frac{v}{rs}. \quad [6]$$

As expected, in a spatial population of constant size N , the likelihood of parallel adaptation increases with the range r of the population and decreases with the speed v at which an adaptive mutation spreads. In the classical Fisher-KPP model for traveling waves [73,74], this speed is given by $\sigma\sqrt{s}$, where σ is the average dispersal distance of an individual per generation [75]. Thus, the probability of soft sweeps depends on the square root of the selection coefficient in this model.

Ralph and Coop [16] extended this approach to populations in arbitrary dimensional spaces by defining a characteristic length $\chi = (v/\lambda\omega(d))^{1/(d+1)}$, where $\omega(d)$ is the area of a sphere of radius one in d dimensions and λ is the intensity per unit area per generation at which adaptive mutations establish in the population. This characteristic length can be interpreted as the average distance travelled by an unobstructed wave until another successful mutation is expected to have arisen within its already covered area (Figure 3C). Consequently, parallel adaptation will be likely if the maximum species range r is larger than χ , whereas hard sweeps should dominate otherwise. In two dimensions, $\omega(2) = \pi$ and $\lambda = \Theta s/(\pi r^2)$, which recovers the condition from Equation 6.

Using estimates for human population density and dispersal rates, Ralph and Coop [16] showed that parallel adaptation could be likely over ranges such as Eurasia once the mutational target size is sufficiently large, for example 1000 bp, approximately the number of coding bases in a human gene.

Note that, in the above scenarios, we only considered situations in which selection acts homogeneously in space. If selection is highly heterogeneous in a patchy environment, migration of the selected allele could be substantially slower than that of unlinked neutral variation. Consider, for example, the evolution of pesticide resistance: if the resistant mutation is strongly deleterious in the absence of pesticides, then the migration of a resistant allele from one patch to another could be slow, as long as it requires survival and reproduction in the habitats where the resistant allele is deleterious.

Understanding $\Theta = 2N\mu$

The theoretical arguments we presented above demonstrate that the key parameter determining the likelihood of soft sweeps is the rate at which adaptive mutations enter the population at a locus. In the Wright–Fisher model, this is given by $\Theta = 2N\mu$, twice the product of population size and the mutation rate towards the adaptive alleles at the locus of interest. This raises two questions: what is a relevant locus and what is the relevant population size in a realistic population?

Definition of a locus

The definition of the relevant genetic locus for a selective sweep is somewhat vague and can range from a single

nucleotide to the whole genome in the case of asexual organisms without recombination (although in this latter situation, soft sweeps are usually interpreted under the notion of ‘clonal interference’ [2,76,77]).

In sexual organisms, a selective sweep generates an effectively linked region around the sweeping mutation. The characteristic size of this region is approximately $s/[r\log(Ns)]$ and, thus, increases with the strength of positive selection and decreases with the recombination rate [8]. Consider the example of lactase persistence, where positive selection was apparently so strong that a genomic segment on the scale of 1 Mb in length has swept through the population without being broken up by recombination during the sweep. In this situation, a second adaptive mutation, even when located hundreds of kilobases away from the first adaptive mutation, could still have led to the observation of a soft sweep at this locus. Note that this second mutation could even have resulted from adaptation to an entirely unrelated selective pressure, although this should only be common when the rate of sweeps is so high that multiple unrelated sweeps can overlap in time and genomic location. Even in *Drosophila*, where sweeps are frequent, we still do not expect this to be common [78].

As a consequence of linkage, stronger positive selection leads to longer loci relevant for selective sweeps. In addition, such longer loci should typically have higher mutation rates towards adaptive alleles, increasing the likelihood of soft sweeps. This brings to light an important difference between the one-locus Wright–Fisher model and a sexual population with recombination: in the Wright–Fisher model, the likelihood of soft sweeps does not strongly depend on the strength of selection. However, in a sexual population, the stronger the positive selection during adaptation, the more likely it should produce soft sweeps.

Note that the relation between strength of selection and likelihood of soft sweeps can be more complex for such loci than under the simple Wright–Fisher model. A longer locus has a larger mutational target size and, thus, potentially a higher rate at which adaptive mutations can occur. However, not all such mutations will necessarily have the same selection coefficients. In addition, we ignore the linkage of adaptive alleles to other fitness-affecting alleles elsewhere in the genome [79]. Thus, the analytical arguments from the Wright–Fisher model no longer hold and understanding of soft sweeps in such realistic situations remains an open topic of investigation.

Population size

The departure from a Wright–Fisher model with constant population size also has profound implications on the other parameter that enters our definition of $\Theta = 2N\mu$: the population size. One can relax many assumptions of the Wright–Fisher model by simply exchanging N with the variance effective population size $N_e = p(1-p)V_{\delta p}$, where $V_{\delta p}$ is the expected variance in population frequency per generation of a neutral allele at frequency p due to drift [80]. However, the strength of drift can vary across time when population size changes, for example during a population bottleneck. Which value of variance N_e is then to be used for estimating the likelihood of soft sweeps in a species?

The answer lies again in the timescales relevant for adaptation. The values of Θ relevant for adaptation are those during the period when the adaptive mutation establishes in the population. Say we were interested in recent adaptation in a human population during the past 500 generations, then only the values of Θ over this time matter. Demographic events that occurred in the more distant past, such as severe bottlenecks occurring during the spread of humans around the globe more than 10 000 years ago [81,82], are not relevant for recent adaptation, even though they could still have profound effects on patterns of neutral variation.

In some species, population size can fluctuate on timescales that are shorter than the time it takes for an adaptive mutation to sweep through the population. In *Drosophila melanogaster*, for instance, population sizes fluctuate by many orders of magnitude between summer and winter [83], and such seasonally driven boom–bust cycles are also likely to be the case for many other insects. To understand the parameters that determine the likelihood of soft sweeps in such cases, consider a highly idealized model of a population of size N_1 that undergoes recurrent severe bottlenecks every ΔT generations, during which its size instantaneously plummets to $N_2 \ll N_1$ (Figure 4A). An adaptive mutation is only likely to survive the next bottleneck if it manages to reach a frequency $x = 1/N_2$ before the next bottleneck occurs so that, on average, at least one copy of the allele is present during the bottleneck. After establishment in the large population, this will take approximately $T_x = (1/s)\log(sN_1/N_2)$ generations of exponential growth. Soft sweeps that emerge during the boom phase are then likely to remain soft throughout the bottleneck only if the second most frequent component of the sweep also had enough time to reach frequency $x = 1/N_2$. This requires that the time between successive bottlenecks, ΔT , is larger than $T_e + T_x$, yielding Equation 7:

$$\Delta T > \frac{1}{\Theta_1 s} + \frac{1}{s} \log\left(\frac{N_1 s}{N_2}\right) \quad [7]$$

Intriguingly, the departure from the simple constant size scenario once again introduces a strong dependence of the likelihood of soft sweeps on the strength of selection. In particular, when population size plummets recurrently on a timescale ΔT , soft sweeps should be the norm for strong mutations, whereas weaker mutations should primarily produce hard sweeps.

Thus, other things being equal, the stronger the selection the more common soft sweeps should be, both because the relevant locus becomes effectively larger, increasing the mutation rate toward adaptive alleles, and because adaptation that starts during a boom phase in a population has time to run its course before the next bust.

Soft sweeps might be the dominant mode of rapid adaptation in many species

Contemporary evolutionary biology is afflicted by an odd dichotomy: experimental evidence suggests that adaptation via selective sweeps is often rapid, involving multiple adaptive mutations that rise in parallel at the same locus, yet population genetic models typically assume mutation-limited scenarios and hard selective sweeps. We argue that

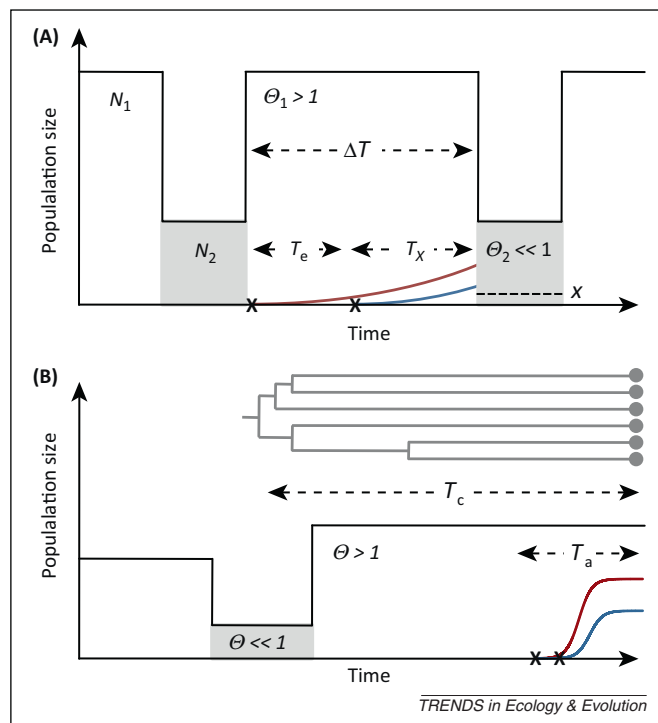


Figure 4. Soft sweeps and demography. **(A)** Probability of soft sweeps under recurrent population bottlenecks. Every ΔT generations, the population size drops from N_1 to $N_2 \ll N_1$. During the boom phase, $\Theta_1 > 1$, but $\Theta_2 \ll 1$ during the bottleneck. Soft sweeps that emerge during a boom phase remain soft throughout the next bottleneck only if at least two mutations reached a frequency $x = 1/N_2$, such that they are likely to survive this bottleneck. **(B)** Difference in variance and coalescence N_e in the presence of a population bottleneck. The timescale of neutral coalescence (T_c) is primarily determined by the time since the bottleneck. Thus, the value of coalescence N_e inferred from the levels of neutral variation can be much smaller than the value of the present-day variance N_e , estimated over the much shorter timescale (T_a) relevant for recent adaptation.

this discrepancy reflects the confusion of two different definitions of the effective population size and that adaptation is not limited by mutation in many species.

As we have discussed above, the key parameter determining whether adaptation is mutation-limited is $\Theta = 2N_e\mu$, twice the product of mutation rate towards the adaptive allele at the relevant locus and the variance effective population size estimated over the timescale relevant for adaptation. Mutation limitation and, consequently, hard sweeps, correspond to scenarios where $\Theta \ll 1$, whereas when Θ is on the order of 1 or larger, adaptation is not limited by mutation and sweeps become soft. This holds true regardless of whether adaptation involves recurrent *de novo* mutations or multiple alleles from the standing genetic variation.

Given that variance N_e is generally difficult to measure over the short timescales relevant for adaptation, one often uses other estimates of N_e based on the relation between the expected level of neutral diversity and expected pairwise coalescence times [84,85]. When population sizes fluctuate fast compared with the timescale of pairwise coalescence, diversity at neutral sites can be used to estimate the harmonic mean of variance N_e over the (generally long) time period until coalescence [86,87]. Importantly, this harmonic mean is dominated by phases where variance N_e is small, even if those phases were short and happened long ago (Figure 4B). Thus, in species with large

census sizes, this ‘diversity’ N_e can be determined primarily by ancient and rare bottlenecks, recurrent selective sweeps, and background selection, even though variance N_e has been large most of the time [10,41,88].

However, whether adaptation is generally mutation-limited in the evolution of a species does not depend on the harmonic mean of variance N_e estimated over long timescales. Adaptation, especially when driven by strong selection, happens fast and adaptive mutations are more likely to arise when the population size is large. The dynamics of adaptation should therefore be determined by how large variance N_e has been during its evolutionary history on average, which is described by the arithmetic mean. The arithmetic mean of variance N_e will always be larger than the harmonic mean and closer to census population sizes.

This difference can explain the observation that soft sweeps are ubiquitous even in species where N_e values inferred from the levels of neutral diversity are low. If the average variance N_e is on the order of the inverse of the mutation rate in these species, then adaptation is not limited by mutation at single sites. Given that many organisms have mutation rates per site between 10^{-8} and 10^{-10} [89], the relevant average variance N_e for adaptation needs to be on the order of 10^8 to 10^{10} for soft sweeps to dominate in these species. This lower bound is in fact conservative because it assumes that all adaptation happens by mutations at a single site in a locus. Larger mutational targets would make soft sweeps more likely and bounds on average variance N_e less stringent. Such values are entirely reasonable for species with large census sizes, especially given that soft sweeps are still common when $\theta \approx 0.1$ [11] and, thus, average variance N_e ten times smaller.

If variance effective population sizes are indeed often this large, then soft sweeps should be the dominant mode of adaptation much of the time. By contrast, hard sweeps should only be common: (i) in consistently small populations; (ii) when adaptation is driven by weak selection in populations of sharply fluctuating size; or (iii) when the mutation rate towards the adaptive allele is extremely low, such as when only a specific combination of mutations is adaptive whereas individual mutations are not [90,91].

The possible prevalence of soft selective sweeps puts pressure on the field of population genetics to develop a more sophisticated understanding of the nonmutation-limited regime. In this regime, the distinction between *de novo* mutations and standing variation becomes blurred because every mutation at every site exists in the population most of the time. Thus, populations should be able to explore the genotype space efficiently and not remain stranded on local fitness peaks for long periods of time [90,92]. Complex, multistep adaptations can arise quickly, with intermediate steps not necessarily reaching high population frequencies [90,93]. Finally, given that genetic drift will be weak most of the time, the patterns and levels of neutral polymorphisms should be primarily determined by the stochastic effects generated by recurrent selective sweeps at closely linked sites, the so-called ‘genetic draft’ [94–98].

Clearly, to arrive at a more comprehensive understanding of the adaptive process, we need to develop better

methods for quantifying soft sweeps in population genomic data, determining their rate and strength, and ultimately identifying the causal adaptive mutations. This task is challenging but holds much promise, given the vast amount of genomic data becoming available and given that many, if not most, cases of adaptation are yet to be discovered.

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