

1 Host-parasite coevolution and its genomic signature

2
3
4 *Dieter Ebert^{1,2,*} & Peter D. Fields¹*

5
6 ¹ University of Basel, Department of Environmental Sciences, Zoology, Vesalgasse 1,
7 4051 Basel, Switzerland

8 and

9
10
11 ² Wissenschaftskolleg zu Berlin, Wallotstrasse 19, 14193 Berlin, Germany

12
13
14 *e-mail: dieter.ebert@unibas.ch

15
16 ORCID IDs:

17 0000-0003-2653-3772 (D.E.)

18 0000-0003-2959-2524 (P.D.F.)

19
20
21 Manuscript accepted for publication in:

22 **Nature Review Genetics, Vol. 21, December 2020**

23 <https://doi.org/10.1038/s41576-020-0269-1>

24
25
26 **Abstract** | Studies in diverse biological systems have indicated that host–parasite
27 coevolution is responsible for the extraordinary genetic diversity seen in some
28 genomic regions, such as major histocompatibility (MHC) genes in jawed vertebrates
29 and resistance genes in plants. This diversity is believed to evolve under balancing
30 selection on hosts by parasites. However, the mechanisms that link the genomic
31 signatures in these regions to the underlying coevolutionary process are only slowly
32 emerging. We still lack a clear picture of the coevolutionary concepts and of the
33 genetic basis of the coevolving phenotypic traits in the interacting antagonists.
34 Emerging genomic tools that provide new options for identifying underlying genes will
35 contribute to a fuller understanding of the coevolutionary process.

40 Introduction

41 Host–parasite coevolution occurs when selection by **parasites** [G] triggers
42 diverse host adaptations that reduce the costs of infection, which in turn prompts
43 parasites to adapt anew to their hosts. This process may be among the most important
44 generators of biological diversity over the past 3.5 billion years ^{1,2}, including the
45 generation and maintenance of genetic diversity within populations and species, and
46 the sharing of certain variants across species boundaries ³⁻⁹. As such, the genomes of
47 organisms would be expected to show signatures of this host–parasite coevolution.
48 Recognizing and characterizing these **genomic signatures** [G] would be expected to
49 lead to a better understanding of how diversity has evolved, and is still evolving, across
50 the tree of life. Furthermore, identifying genomic signatures of coevolution can help to
51 narrow down the loci under selection and provide mechanistic insights into host–
52 parasite interactions.

53 Antagonistic coevolution has been defined as reciprocal selection between two
54 closely interacting species ^{10,11}. This definition focuses on the phenotypic traits of the
55 coevolving antagonists that negatively influence each other. It specifies that it is the
56 traits responsible for the interaction and their underlying genes that coevolve—not
57 the species ^{12,13}. It also sets antagonistic coevolution apart from scenarios of host–
58 parasite co-association and the resulting long-term patterns of co-speciation or co-
59 cladogenesis, which do not require adaptive processes ¹⁴.

60 The process of antagonistic coevolution can be described by two types of
61 model: models of specific coevolution, in which one host and one parasite species
62 interact (sometimes called pairwise coevolution); and models of unspecific
63 coevolution, better known as diffuse coevolution ¹⁰, in which multiple hosts and/or
64 parasite species contribute to the process. Models of specific coevolution can be
65 further grouped into **selective sweep** [G] models, in which novel variants are selected
66 for and rise to high frequencies, and balancing selection models, in which alternative
67 variants at specific loci fluctuate in frequency over time ^{11,15}. The different modes of
68 action underlying these models create different signatures in the genomes of the
69 antagonists. Thus, observation of a particular signature in a genome allows (within
70 limits) conclusions to be drawn about the evolutionary mechanism producing them
71 (Table 1).

72 A genomic perspective of coevolution considers the entire region around the
73 coevolving genes, because making use of the additional information present in the
74 flanking sites strongly increases the power of the analysis to detect genomic
75 signatures. Despite being functionally independent, sites physically tightly-linked to the
76 coevolving genes are influenced by **linked selection** [G], which causes **genetic**
77 **hitchhiking** [G], and, unlike unlinked sites, their fate is determined by the dynamics of
78 the selected genes and the rate of recombination among them ^{16,17}. Thus, genomic
79 signatures of coevolution can be detected in populations by comparing patterns of

80 genetic variation in these regions with the patterns in the genomic background that
81 presumably evolved by neutral evolutionary processes. The genomic signatures of
82 models of specific coevolution have received much attention, resulting in a good
83 picture of the expected patterns in population samples of genomes. By contrast, the
84 genomic signatures expected under diffuse coevolution have not yet been
85 determined; nevertheless, some heuristics exist.

86 In this Review, we address conceptual issues concerning host–parasite
87 coevolution and how they manifest in the genomes of the antagonists. We highlight
88 the differences between specific and diffuse coevolution and point out the role of
89 spatial population structure in shaping genomic signatures. Using these concepts, we
90 discuss the evolution of trans-species polymorphisms. Finally, we address new
91 **cogenomic [G]** methods that allow coevolving loci in hosts and parasites to be directly
92 pinpointed and what we can learn from these recent developments.

93

94 **Models of specific host–parasite coevolution**

95 The two models of specific coevolution, selective sweep selection and balancing
96 selection, have different effects on genetic variation. Selective sweep coevolution is
97 arguably one of the most important generators of macroevolutionary patterns, and
98 explains differences in immune systems among lineages, drives speciation and is
99 implicated in some of the major transitions in evolution^{2,18-20}. However, it fails to
100 explain the extraordinary genetic diversity observed at some host resistance loci. By
101 contrast, coevolution by **balancing selection [G]** is best known for its potential to
102 maintain high levels of genetic diversity within populations and species. The theory
103 states that hosts and parasites undergo continuous antagonistic coevolution, often
104 referred to as Red Queen dynamics or trench warfare, which results in a balance of
105 different variants at loci related to host defence and parasite offence²¹⁻²³. Recent
106 studies have revealed that there are more genomic regions in plants and animals that
107 undergo balancing selection than previously thought^{8,24-29}.

108 The study of coevolution by analysing patterns of genetic variation has a long
109 tradition, but has a number of limitations that make it hard to reach strong conclusions
110 about underlying evolutionary processes. For example, genomic studies on
111 coevolution are mostly performed separately in host and parasite genomes (but see
112 ‘Joint analysis of host and parasite genomes’). Thus, a given finding may be difficult to
113 attribute to host–parasite coevolution or to other processes, because the
114 bioinformatic and population genetic approaches that are used to identify genomic
115 regions with characteristics of selective sweeps or balancing selection do not identify
116 the cause of selection — antagonistic interactions are only one of many possibilities.
117 Additional efforts are needed to differentiate between these options. This issue is
118 particularly pronounced for balancing selection, as different mechanisms (such as
119 **overdominance [G], local adaptation [G], direct negative frequency dependent**

120 selection [G] (NFDS) and indirect negative frequency dependent selection [G]) can all
121 produce genomic signatures of balancing selection. However, only indirect NFDS is
122 associated with antagonistic coevolution, and so distinguishing among these
123 mechanisms is crucial.

124

125 *[H2] Selective sweep coevolution.* A selective sweep describes the rapid increase in
126 frequency of a beneficial variant, such as a novel mutation. In the extreme case, the
127 variant will reach fixation, replacing alternative variants. Adaptive evolution in this
128 form is considered to be a dominant driver of protein adaptation and species
129 divergence^{30,31}. During host–parasite coevolution, both antagonists may experience
130 sweeps at loci that play a functional role in their interaction (FIG. 1)^{11,15,32–35}. For
131 coevolution to occur, sweeps do not need to alternate between hosts and parasites.
132 Variants at multiple sites may even spread at the same time, as long as they are
133 decoupled by recombination, as otherwise selective interference [G] takes place³⁶
134^{33,34}.

135 A selective sweep leaves a characteristic valley of locally reduced genetic
136 variation in population samples of genomes. These diversity valleys are formed
137 because variants that are in proximity to the sweeping variant hitchhike along and
138 replace alternative variants, which results in a corresponding pattern of increased
139 linkage disequilibrium [G]³⁷. Such valleys allow us to detect the approximate position
140 of the beneficial mutant, but the genomic signatures are rarely clear and lose their
141 distinct features over time as recombination breaks down associations and as new
142 mutations arise. Thus, ancient selective sweeps are no longer detectable, although
143 they leave their traces in signatures of positive selection that result in patterns of
144 divergence among lineages²⁰. Likewise, sweeping mutants in the initial phase of a
145 sweep are hard to detect, as the reduction in local genetic diversity is not yet
146 pronounced (FIG. 2). Additionally, each sweep event has a different history: each one
147 is associated with a different selection coefficient [G] and may occur at a different
148 location in the genome and possibly experience different local recombination rates³⁸.
149 Finally, signatures of sweeps strongly depend on the initial variation on which selection
150 acted³⁷.

151 Variants spreading to fixation in coevolving hosts and parasites may be
152 functionally independent, other than that they each provide a benefit to their carrier
153 and are disadvantageous to the antagonist²³. However, when particular host or
154 parasite proteins interact, selective sweep coevolution can occur through alternating
155 changes to functionally important segments of these proteins. For example, the
156 restriction factor *TRIM5*, an antiviral protein in Old World primates seems to have
157 coevolved in tight interaction with capsid proteins in lentiviruses³⁹.

158 Due to their high rate of evolution, parasites are among the fastest changing
159 selecting agents host organisms experience. Therefore, the speed with which a host

160 population can respond to a new variant of a parasite is important. However, selective
161 sweep evolution from a novel mutation to fixation (that is, a hard sweep) is a rather
162 slow process. In particular, the initial increase in allele frequencies is slow: in a large
163 population, a novel mutation with a 10 % fitness advantage takes about 200
164 generations to reach a frequency of 10 %, but only about 15 generations more to
165 reach 50 %^{40,41}. Thus, sweeps starting from standing genetic variation (known as soft
166 sweeps) can occur faster than hard sweeps, as the initial allele frequencies are higher.
167 Indeed, evidence is accumulating that many sweeps observed in natural populations
168 are soft sweeps⁴¹⁻⁴³. Interestingly, standing genetic variation is often high for loci
169 related to host–parasite interactions and thus can drive faster sweeps. Although it is
170 not fully known why this is the case, the two-speed genome model⁴⁴⁻⁴⁶ attempts to
171 explain this paradox by suggesting that specific regions in host and parasite genomes
172 have strongly elevated mutation rates, constantly re-fuelling standing genetic
173 variation, whereas housekeeping genes have lower mutation rates and evolve by
174 purifying selection.

175

176 *Coevolution by balancing selection.* The Red Queen model of balancing selection was
177 introduced by Clarke⁴⁷ to explain unexpectedly high genetic diversity within
178 populations. It is based on the assumption of strong host–genotype by parasite–
179 genotype interactions⁴⁸⁻⁵⁰ (FIG. 3a), which has been finding increasing empirical
180 support⁵¹⁻⁵³. As opposed to the selective sweep model, in which mutant alleles have a
181 general advantage over the wild type alleles, genetic variants under balancing
182 selection provide an advantage only in specific relation to a corresponding genetic
183 variant in the antagonist. These host and parasite genes are functionally coupled,
184 forming the driving force for reciprocal selection^{22,54-56}. According to the Red Queen
185 model, a parasite allele will increase in frequency when the host allele that allows it to
186 infect is common; as hosts carrying this allele succumb to increased parasitism, their
187 numbers will decline, and—with a time lag—so too do parasite genotypes that depend
188 on these particular host genotypes, that is, the loci in the host and the parasite are
189 under NFDS. As parasites with particular infectivity alleles track corresponding host
190 alleles, cyclical dynamics of host and parasite alleles might arise, with parasite cycles
191 lagging behind the corresponding host allele cycles^{57,58}. NFDS reduces the likelihood
192 that alleles go extinct by chance because rare alleles gain an advantage; thus, genetic
193 polymorphisms in hosts and parasites can be maintained, and the frequencies of the
194 functional variants at the involved loci are balanced over time (FIG. 3b). In this form of
195 balancing selection, an allele’s selection coefficient does not depend directly on its
196 own frequency, but rather depends on the frequencies of specific alleles in the
197 antagonist. Therefore, it is called indirect NFDS²¹.

198 Relative to the rest of the genome, genomic regions containing genes under
199 balancing selection are expected to show high genetic diversity and a more even

200 frequency spectrum of variants, resulting in positive Tajima's D [G] values^{16,26,59,60}
201 (Table 2). Scans of host genomes have revealed ample evidence of regions under
202 balancing selection, with the most well-known being the MHC class 1 and 2 genes of
203 jawed vertebrates and the ABO blood group system of higher primates, but also other
204 regions of vertebrate genomes related to immune function^{25-28,61-64}. Scans in plants
205 and invertebrates have also revealed balancing selection in diverse genomic regions,
206 again often regions associated with immune function^{8,65-71}. However, current
207 bioinformatic and population genetic methods may still miss many regions of interest.
208 To produce a recognizable genomic signature, selection should be strong and should
209 have acted for a sufficiently long time period (Fig. 2). Regions around sites under
210 selection usually show strong local linkage disequilibrium, thus local recombination
211 rate will also affect the ability to detect signatures. Ideally, the recombination rate
212 around the selected loci should be low to allow linked SNPs to hitchhike; high
213 recombination rates would reduce the region to the very site under balancing
214 selection itself^{16,21}. However, some level of recombination is helpful for the detection
215 of the sites under selection, as otherwise the entire region will be inherited as one
216 linkage block.

217 Less attention has been given to balancing selection in parasites than in hosts.
218 Mapping parasite loci is more difficult because parasites often cannot be cultured and
219 phenotyped outside the host. Furthermore, they often have no or infrequent genetic
220 recombination, have extreme population structures and demography, carry extra
221 chromosomal genetic elements (such as plasmids), and have substantial divergence in
222 gene content as a result of horizontal gene transfer: each of these features may
223 produce genomic signatures that make it more difficult to identify the signature of
224 balancing selection^{72,73}. New bioinformatic tools for genome analysis are now being
225 developed that address some of these challenges (see Table 3 in ref⁷³), and
226 theoretical analyses suggest that balancing selection might show stronger genomic
227 signatures in parasites than hosts²¹. Examples of loci likely under balancing selection
228 have been described for a range of parasites⁷⁴⁻⁷⁹. For West Nile Virus in mosquitos and
229 Deformed Wing Virus in honey bees it has been suggested that genetic diversity is
230 maintained by selection from RNA interference (RNAi) in the host^{78,80,81}. Interestingly,
231 the genes for RNAi in *Drosophila* were suggested to evolve by very high rates of
232 positive selection with evidence for recent selective sweeps⁸². Two closely-related
233 human pathogenic bacteria, *Staphylococcus aureus* and *S. epidermidis*^{74,75}, show
234 evidence for balanced polymorphisms, but for different genes. The hrpA gene of the
235 plant pathogen *Pseudomonas syringae*, which encodes part of the type III secretion
236 system, was found to be under balancing selection⁷⁶. In addition, a rare example of a
237 balanced polymorphism involving structural variation, which determines the presence
238 or absence of alternative pathogenicity islands, was described in *Pseudomonas*
239 *viridiflava*, a plant pathogen often found on *Arabidopsis*⁸³. Interestingly, several of

240 these bacteria are known for their wide host range, hinting that diffuse coevolution
241 may contribute to the observed patterns (see ‘Diffuse coevolution’ section). Sites with
242 high genetic diversity and polymorphisms have also been described in human malaria
243 parasites ^{84,85}, although the signature of balancing selection here may be confounded
244 with processes resulting from acquired immunity. Finally, in the planktonic crustacean
245 *Daphnia magna* and its highly specific bacterial parasite *Pasteuria ramosa*, evidence
246 for balancing selection comes from both the infectivity loci in the bacteria ⁷⁹ and from
247 the resistance loci in the host ⁵³.

248 The genomic signatures of coevolution in hosts and parasites cannot be
249 expected to resemble each other, as the two antagonists experience selection in very
250 different ways. For parasites, the fitness difference between infecting and failing to
251 infect a host is very large, whereas the lifetime fitness difference between hosts that
252 resist a parasite attack or not is likely much smaller ⁸⁶. In addition, some hosts may not
253 encounter the parasite during their lifetime, reducing selection for resistance even
254 further. Hosts and parasites also may have different generation times, recombination
255 rates, effective population sizes, ploidy-levels, mutation rates, population structures
256 and demography, to name just a few dissimilarities. Therefore, the footprint of
257 coevolution and the genomic signatures of the two coevolving antagonists may look
258 very different and may sometimes only be detectable in one of the antagonists ²¹.

259 Initially, population genetic modelling of coevolution by NFDS was performed
260 for infinite population sizes, which strongly reduces the chance of extinction of host
261 and parasite variants. Real populations, however, are of finite size and subject to
262 stochastic effects, so functional variants may be lost due to **genetic drift** [G], bringing–
263 in the most extreme case–coevolution to a halt ^{21,87-89}. These effects can substantially
264 blur the genomic signatures of coevolution in natural populations, making their
265 interpretation more difficult ²¹. However, spatial structures of host populations can
266 strongly reduce these stochastic events, as long as sub-populations are not evolving in
267 synchrony ⁹⁰⁻⁹².

268
269 *Specific coevolution and spatial structure* Simple models of host–parasite
270 coevolution generally focus on evolutionary dynamics in a single, large, **panmictic** [G]
271 population. However, populations are almost always spatially distributed, with an
272 extended geographic structure and gene flow among populations. This spatial
273 structuring profoundly influences coevolution and divergence ^{58,90,91,93-97} and
274 accounting for it can strongly refine population genomic analyses ^{33,73,98}.

275 Research on subdivided populations has examined how coevolution influences
276 the spread of variants related to host–parasite interactions, thereby contributing to
277 patterns of genetic variation on a species-level scale (rather than a population-level
278 scale) ^{96,99,100}. In a selective sweep scenario, strong gene flow allows globally beneficial
279 mutations to spread quickly from population to population, whereas weak gene flow

280 leads to population divergence^{37,101}. More complex patterns of divergence may arise,
281 for example when gene flow is rare and unbalanced among subpopulations, when it
282 differs for the two antagonists, and when populations adapt locally to other
283 environmental factors¹⁰². Genomic patterns of such processes are hard to study,
284 especially if the evolving genes are not known beforehand^{37,101}.

285 By contrast, under balancing selection, gene flow enables alleles to persist in
286 the overall gene pool much longer, as it can bring locally extinct alleles back into the
287 population (FIG. 3b)^{60,93,100,103,104}. In addition, as immigrating alleles are likely to be
288 rare upon their arrival, they are expected to have an advantage and experience a
289 lower rate of extinction and a higher likelihood of spreading locally^{104,105}, thereby
290 increasing their effective migration rate compared to neutral genetic variants. The
291 number of alleles maintained on a species level thus rises accordingly, and the loci
292 under NFDS are expected to show less differentiation among populations than neutral
293 loci that do not benefit from NFDS^{79,106}. This difference is seen when comparing
294 isolation-by-distance (IBD) patterns for these two groups of loci: whereas neutral loci
295 show increasing differentiation (visible, for example, as an increased fixation index [G]
296 (F_{ST})) with increasing distance among populations, loci involved in NFDS show no or a
297 much weaker pattern of IBD¹⁰⁷, which is the opposite of what local positive selection
298 would produce^{104,108,109}. However, while the immigrant advantage reduces
299 differentiation at the loci under NFDS among populations over the long term, it may
300 drive the divergence of neighbouring populations in the short term, for example if
301 strong selection occurs at a given gene but for different variants. Such patterns have
302 been observed in some population genetic studies of resistance genes^{100,106,110}. Thus,
303 populations may show more variable pairwise F_{ST} at sites under NFDS than at neutral
304 sites.

305 Although limited gene flow causes populations to show patterns of IBD^{111,112},
306 the perspective described above does not include common ecological and biological
307 circumstances that may influence gene flow, such as spatially divergent selection,
308 population size and genetic drift, metapopulation and source-sink population
309 structure, and historic events. Parasites may differ in local abundance and may not be
310 present in every host population. Thus, the propensity for host–parasite coevolution
311 may vary in space and time. Host genes involved in host–parasite interactions may
312 become neutral in the absence of the parasite, or even detrimental if there are costs
313 of resistance. The combined dynamics of these and other evolutionary and ecological
314 processes that influence spatial and temporal variation have been described as the
315 geographic mosaic of coevolution, with hot spots showing strong and rapid
316 coevolutionary dynamics, and cold-spots marked by slow or no coevolution^{113,114}. For
317 any given system, the greater the proportion of coevolutionary cold spots, the weaker
318 the overall signature of balancing selection. However, even at cold spots, one may find
319 elevated genetic diversity at loci that are under balancing selection in hot spots,

320 because gene flow from hot spots may prevent extinction of variants in cold spots.
321 Species-wide, the overall genomic signature of balancing selection may, therefore, be
322 relatively insensitive to the geographic mosaic of coevolution. However, this will
323 certainly depend on the interplay of migration, genetic drift and selection.
324

325 **Trans-species-polymorphism**

326 Because alleles under balancing selection are less likely to be lost from a gene-
327 pool, they segregate for longer time periods than genes undergoing neutral evolution
328 or **directional selection** [G]¹⁶. This theory can be tested using coalescence approaches
329 that compare regions under balancing selection with the genomic background³⁸. In
330 extreme cases, ancient identical-by-descent genetic variants at polymorphic sites may
331 have pre-existed even before the last speciation event, so that these variants are
332 shared across closely related species (FIG. 4a). This trans-species-polymorphism (TSP)
333 is indicated when haplotypes from the two species cluster by allele and not by species
334 (FIG. 4b)^{16,115}, as expected for neutrally evolving alleles. In the context of host-parasite
335 interactions, the signal of TSP is very clear in MHC genes, in the ABO blood group
336 system of higher primates and in several other vertebrate loci^{25,29,64,115-117}. It has also
337 been observed in plants^{8,118,119}, but rarely in invertebrates⁶⁸. Genomic regions with
338 evidence of TSP are typically enriched with immune function genes^{6,8,26,97,105,119,120},
339 suggesting that host–parasite coevolution might be the driving force behind a genomic
340 signature of TSP^{20,120,121}. Experimental data further support a link between TSP and
341 parasitism^{105,122}.

342 Given the large time scale involved in TSP, the impact of the coevolving genes
343 on the immediate genomic neighbourhood of the selected sites will be reduced,
344 because recombination decouples selected sites from linked neutral sites^{16,123}.
345 Therefore, long-term polymorphisms of single SNPs are difficult to detect by analysing
346 genomic signatures alone. This can be seen with the SNP in the human haemoglobin
347 gene responsible for the balanced sickle-cell anaemia–malaria resistance
348 polymorphism in large parts of Africa. Long-term balancing selection can best be
349 detected when the balanced haplotypes do not recombine (for example, if they are
350 located in inverted regions of the genome or are part of a **supergene** [G]), or if
351 multiple polymorphisms are together under selection, so that they maintain linkage
352 disequilibrium across polymorphic sites^{25, 53, 123}.

353

354 **Diffuse coevolution**

355 The two models of specific coevolution described above assume that the relevant
356 interactions are between one host and one parasite species. However, interactions
357 may include multiple parasites and/or multiple hosts. This scenario is known as diffuse
358 coevolution¹⁰. Diffuse coevolution refers particularly to the interactions of **functional**
359 **guilds** [G], such as diverse species of herbivores and their plant hosts or parasites and

360 hosts. In its simplest form, coevolution is diffuse when a host trait and the underlying
361 genes in the genome evolve in response to at least two parasites, or a parasite trait
362 evolves in response to selection caused by more than one host. In such cases,
363 coevolution is influenced by a combination of frequency dependence (according to the
364 genetic composition of the parasite populations) and density dependence (according
365 to the abundance of the parasite species). Furthermore, the parasite composition may
366 differ in different sub-populations of the host and may change over time (owing to
367 diversifying selection and/or spatio-temporal dynamics)¹²⁴⁻¹²⁶. The more hosts or
368 parasites that participate, the more complex the evolution of the participants will be
369 and the harder it will be to predict the genomic signature of the diffusely evolving
370 regions¹³.

371 The need for consideration of diffuse coevolution is underlined by examples of
372 host loci that interact with multiple parasites. Human variants of CCR5, TRIM5 α and
373 APOBEC3 interact with HIV-1, but were suggested to have been under positive
374 selection by other viruses with which they interacted in the past¹²⁷⁻¹²⁹. The tomato **R-**
375 **gene** [G] *cf2* confers resistance to the parasitic nematode *Globodera rostochiensis* and
376 to the fungus *Cladosporium fulvum*¹³⁰. Parasites from three kingdoms interact in part
377 with the same proteins in *Arabidopsis*¹³¹. The MHC region of diverse vertebrates are
378 well known for their interactions with many parasites^{125,126,132,133}.

379 The converse, a parasite interacting with more than one host, seems also to be
380 widespread. Sympatric species of sticklebacks are infected by the same parasites,
381 which interact with the host's MHC¹¹⁶. The pathogenic bacteria *Staphylococcus*
382 *aureus*, *Pseudomonas viridiflava* and *Pseudomonas syringae* are known to infect
383 various host species with overlapping genetic mechanisms^{74,76,134}. Finally, hosts may
384 even exchange immune genes to fight parasites: for example, related *Arabidopsis*
385 species have been shown to exchange resistance genes through hybridization and
386 introgression¹³⁵. Likewise, parasites may exchange coevolving genes via horizontal
387 gene transfer to overcome host defenses¹³⁶.

388 Coevolution may become even more diffuse when the interactions of various
389 hosts and parasite species show spatio-temporal dynamics. Coevolution in such multi-
390 species scenarios is based on variable interactions between changing communities of
391 hosts and parasites. Thus, the signatures observed in the host and parasite genomes
392 cannot be attributed to one ecological setting, but result from a history with diverse
393 settings with different interactions in host and parasite communities. Each setting may
394 result in a different evolutionary trajectory and may well include temporal phases of
395 more narrowly defined coevolution by NFDS or selective sweeps between pairs of
396 antagonists¹³⁷. The signatures of these phases of specific coevolution will not be
397 recognizable in the genome, unless the ecological setting in which they occur are
398 sufficiently stable for a given period of time.

399 Models of single hosts coevolving with multiple parasites support the idea that
400 long term maintenance of genetic polymorphisms is possible^{7,48,59,124,138,139}. Empirical
401 data from systems likely under diffuse coevolution are consistent with this idea,
402 leading to the widely held belief that the complexity of multi-species interactions
403 maintains genetic diversity, including TSP^{92,116,125-127,140-142}. Thus, diffuse coevolution
404 offers an explanation for the "missing antagonist problem" that typifies TSP studies,
405 which typically lack knowledge of the parasite that coevolved with the host in the past
406¹²⁰. If TSP is caused by diffuse coevolution there is no single coevolving parasite, but an
407 association with a changing pool of parasite species over time and space. Indeed, most
408 of the associations of the human MHC with parasites (such as HIV, West Nile virus,
409 dengue, hepatitis B, hepatitis C, tuberculosis and leprosy¹³³) are believed to be rather
410 young—much younger than 5 million years, the approximate date of the last common
411 human–chimpanzee ancestor for which TSP at the MHC was observed^{25,29,116}. Thus,
412 the parasites we see interacting with humans now may not be the same as those that
413 interacted with us some million years ago. Malaria, which is a parasite specific to
414 humans, may be an exception. Recent work suggests that human–malaria coevolution
415 could be as old as the human split from our closest living relatives¹⁴³⁻¹⁴⁶. However, the
416 hypothesis that malaria is a missing antagonist, explaining TSP in humans, requires
417 further investigation.

418 To further our understanding of TSP it will be necessary to distinguish between
419 cases of specific long-term coevolution and diffuse coevolution. Although
420 demonstrating that TSP is the consequence of ancient specific coevolution is
421 challenging, we can make testable predictions. In the strictest case of specific
422 coevolution by balancing selection, the aim would be to demonstrate that two closely
423 related host species with evidence of TSP are parasitized by two closely related species
424 of coevolving parasites (FIG. 4a) and that the genes in question play a functional role in
425 the interaction of the two pairs of antagonists (FIG. 3a). In such a scenario, an overlay
426 of long-term Red-Queen dynamics with the co-speciation of the host and the parasite
427 might be observed¹⁴(FIG. 4b). Such a strict set of conditions may not be very likely, but
428 the requirements may be relaxed to accommodate ecological, historical,
429 epidemiological and biogeographical features^{116,138}.

430

431 **Joint analysis of host and parasite genomes**

432 Unlike the approaches described above, emerging **cogenomic** [G] methods
433 jointly analyse the genomes of hosts and those of their parasites. These methods focus
434 on the genes responsible for the phenotypic interactions of the antagonists and allow
435 subsequent analysis of their genomic signatures^{72,147-149}.

436

437 *[H2] Identifying the genes involved in host–parasite interactions.* A cogenomics
438 approach to identify the genes that directly interact with each other in hosts and

439 parasites is best illustrated with a host–parasite matrix that shows the functional
440 specificity of variants of the antagonists that interact with each other to produce
441 phenotypes—disease (compatible) and resistance (incompatible) (FIG. 3a). Although
442 many different matrices have been proposed ¹⁵⁰⁻¹⁵⁵, we still do not know much about
443 them in natural host-parasite systems. Studies of interacting host–parasite genes have
444 a long tradition in plant–pathogen systems, starting with H.H. Flor’s gene-for-gene
445 system for flax and one of its rust pathogens ¹⁵⁶ (for reviews see: ^{22,91}), but only a few
446 examples have so far been confirmed for animal systems ^{52,53}. It is widely believed that
447 specific interaction matrices like these underlie coevolution by NFDS and thus are
448 responsible for maintaining genetic diversity. Finding the genes underlying such
449 matrices is still cumbersome and time consuming, but this may change with the
450 development of cogenomics approaches that detect interspecies (or intergenomic)
451 linkage-disequilibrium (iLD). If the only polymorphism in a population that explains
452 variation in infection success is the interaction between the host’s A-locus and the
453 parasite’s B-locus, then infected individuals could only ever carry the host–parasite
454 allele combinations A/B and a/b. Other combinations (A/b and a/B) would not produce
455 infections (FIG. 3a). This non-random association between host and parasite alleles
456 produces iLD because the phenotype (infection) depends on the combination of
457 genetic variants in different species. The statistical signal of iLD is only seen at the
458 interacting loci, not at sites in the genetic background, as long as recombination
459 efficiently decouples selected loci from non-selected loci (FIG. 5a). Using iLD is a
460 powerful tool to detect interacting genes for a number of reasons. First, its presence
461 indicates that host and parasite individuals carry alleles related to each other in a
462 fitness context, that is, for a trait (infection) likely under selection. Second, it allows the
463 interacting loci of both antagonists to be pinpointed. Third, it can be used even when
464 only samples of infected hosts (such as infected patients) are available. Fourth, it is
465 free from assumptions about the shape of the interaction matrix, which is important,
466 as we currently have little understanding of how the matrix is structured. Finally, the
467 method is not limited to model species, as it does not require previous knowledge
468 about the system.

469 However, statistical limitations are apparent. The method works best when the
470 alleles in question have intermediate frequencies and when the number of tests to be
471 conducted (every host polymorphism is tested in relation to every parasite
472 polymorphism) is not too large relative to the sample size: as the number of tests, and
473 therefore false positives, increases with the product of the numbers of polymorphisms
474 in both antagonists, larger sample sizes are required for antagonists with large
475 genomes. In addition to these limitations, other factors can reduce the power to
476 detect significant iLD. For example, if genes involved in iLD have epistatic (that is, non-
477 additive) interactions with other loci in the same genome, the statistical power to
478 detect them is reduced. Furthermore, simulations of different coevolutionary models

479 showed that the inferential power of iLD will vary over time due to the underlying
480 dynamics of a host-parasite interaction matrix, and relevant loci will be more
481 identifiable if coevolution occurs by NFDS than if by selective sweep coevolution ¹⁵⁷.
482 Finally, multiple infections (that is, more than one parasite genotype being involved in
483 the infection of individual hosts) may cause problems in the analysis. Multiple
484 infections of humans and animals are believed to be very common ¹⁵⁸.

485 A number of pioneering cogenomics studies have been performed with human
486 patients infected with different parasites. In genome-to-genome analyses, iLD was
487 used to test for associations between SNPs in humans with SNPs in HIV-1 ¹⁴⁷, hepatitis
488 C ¹⁵⁹ and *Streptococcus pneumoniae* ¹⁶⁰. These studies were able to identify previously
489 unknown interacting genes in the host and parasite, with the virus studies being more
490 powerful because they have smaller genomes than *Streptococcus* and require fewer
491 tests. This method is currently being further developed to increase its sensitivity and to
492 take population stratification into account ^{161,162}. A modified version of this method
493 was used to test for associations of SNPs in human candidate genes with phylogenetic
494 lineages of *Mycobacterium tuberculosis*, revealing a host SNP – parasite lineage
495 association ¹⁶³.

496 Another cogenomics method, Analysis with a Two-Organism Mixed Model
497 (ATOMM), also aims to find interacting loci in hosts and parasites (FIG. 5b)¹⁶⁴. This
498 method requires experimental data for phenotypes, such as infection and disease
499 symptoms, from all possible combinations of host and parasite genotypes. It allows
500 researchers to map these phenotypes to host and parasite genomes, while accounting
501 for confounding factors such as phylogenetic or spatial structure.

502 Although not yet routinely used, cogenomic methods offer exciting possibilities
503 for future studies of host–parasite interactions. The new pairs of loci identified with
504 these methods can then be analysed for their genomic signatures, allowing the picture
505 of the signature of coevolution in host and parasite genomes to be fine-tuned. The use
506 of the genome-to-genome method with non-model organisms will allow the study of
507 systems previously out of the reach of coevolutionary research.

508
509 *[H2] Using cogenomics to infer signatures of the coevolutionary process.* As of yet, we
510 do not have access to analytical derivations that determine exactly how different
511 coevolutionary processes affect allele frequencies at loci across host–parasite
512 genomes. However, we can simulate much of the biological complexity that arises as a
513 result of the coevolutionary process and explore how genomic signatures might differ
514 under particular scenarios ¹⁵⁷ (BOX 1). For example, one approach simulates
515 coevolutionary dynamics to generate allele frequency expectations, which can then be
516 used to look back in time to determine the coalescent properties that might generate
517 these expected allele frequencies at loci in host and parasite genomes ²¹. This general
518 idea has been adapted into an inferential framework using **Approximate Bayesian**

519 **Computation** [G] (ABC) ¹⁴⁹ for population genetic simulations of different
520 coevolutionary scenarios. The results of these simulations can be summarized by both
521 traditional population genetic summaries (Table 2) as well as novel summaries of the
522 host and parasite allele frequency spectra concurrently, which are then statistically
523 compared with polymorphism data obtained from host and parasite genomes to
524 determine their similarity. This approach enables the identification of loci that underlie
525 the coevolutionary process in both host and parasite genomes simultaneously, as well
526 as identifying important eco-evolutionary parameters such as the cost of infectivity.
527 The method can be applied to genome wide polymorphism data gathered from
528 controlled laboratory experiments or from natural populations. It may be used either
529 in concert with the iLD approach or entirely independently, as the false positive rate
530 for detecting genomic regions undergoing coevolution is rather low.

531

532 **Conclusions and future perspectives**

533 In a world of rapidly increasing transportation and migration, accelerating climate
534 change, de- and reforestation, altered agricultural and food handling practises, and
535 increasing human and livestock densities, host–parasite contact rates are already very
536 high and are only expected to increase, resulting in more intense reciprocal selection.
537 These changes affect many naturally coevolving systems in which humans have a
538 vested interest, such as malaria–mosquito, virus (including dengue, zika, West-Nile)–
539 mosquito, *Borrelia*–ticks, and *Mycoplasma*–house finch, as well as parasites believed
540 to coevolve with humans ^{163,165-167}. Although we still have a limited understanding of
541 how coevolution works in natural populations, the knowledge derived from the state-
542 of-the-art population genomic methodology described here can guide and inform
543 coevolutionary research and experimentation. We are now in a position to approach
544 new questions and gain a new perspective on old ones.

545 With few exceptions, we know little about the role of structural variation, such
546 as copy number variation and inversions, in the coevolutionary process. Copy number
547 variation, which is well known in the MHC, can be maintained by balancing selection
548 (reviewed in ¹²⁵). Supergenes have also been shown to play an important part in
549 balanced polymorphisms in various traits ¹⁶⁸, and have recently been suggested to
550 have a decisive role in host–parasite coevolution ⁵³. Although it is still cumbersome to
551 identify structural polymorphisms in large samples, improved methods will certainly
552 make this an important aspect of coevolutionary research.

553 A broad survey of the literature on coevolution would suggest a taxon-specific
554 division between plant systems, centred on gene-for-gene infection models, and
555 animal systems, focused on matching-allele-models without invoking costs. It is
556 currently not clear if this division reflects a bias in our research efforts or has a
557 biological basis. Now, cogenomic methods will enable us to move beyond a few model
558 systems and rapidly collect more data on the type of infection matrix that

559 predominates in natural populations. This information is particularly required for
560 animal systems and non-agricultural plant systems, for which there are only a few
561 examples of interaction matrices. Understanding the costs of resistance will need
562 fitness assays in the presence and absence of the parasite.

563 Another challenge is to understand how genes within the same genome
564 interact to modify the dynamic with genes in the antagonist. Epistasis has long been
565 thought to be important in host-parasite interactions⁴⁸, and epistasis between host
566 resistance loci has been proposed for a number of systems^{22,52,53,133,169}. However, the
567 generality and importance of these observations for host–parasite coevolution is not
568 yet clear. Epistasis is central to the theory of coevolution and of the evolution and
569 maintenance of genetic recombination. Genetic recombination works to speed up
570 evolutionary responses under antagonistic coevolution only if the recombining loci
571 show epistatic interactions^{48,170,171} because epistasis may create negative linkage
572 disequilibrium among resistance alleles.

573 Knowing which genes are functionally linked and what epistatic relationships
574 exist would allow us to predict how reciprocal selection acts, whether specific or
575 diffuse coevolution occurs, and what traces we might expect at the genome level.
576 After decades of coevolution research, we still do not have examples of temporal
577 dynamics allele frequencies and their associated phenotypes in either hosts or
578 parasites, or an understanding of how the corresponding genomic sites in the
579 antagonists are functionally linked. However, recent progress in the field promises
580 change in the near future. Finally, our understanding of the genomic signatures of
581 coevolution is still largely correlational. Rigorous experimental work, including
582 experimental evolution studies, can help us scrutinize the evidence for coevolution,
583 link it to genomic signatures, and test specific model predictions^{56,172-174}.

584
585
586

587 **References**

588

589 1 Majerus, M., Amos, W. & Hurst, G. *Evolution: The Four Billion Year War* (Addison,
590 Wesley Longman, 1996).

591

592 2 Jack, R. & Du Pasquier, L. *Evolutionary Concepts in Immunology*. (Springer Nature
593 Switzerland, 2019).

594

595 3 Fumagalli, M. *et al.* Signatures of Environmental Genetic Adaptation Pinpoint
596 Pathogens as the Main Selective Pressure through Human Evolution. *PLoS Genet.* **7**,
597 doi:10.1371/journal.pgen.1002355 (2011).

598

599 4 Karasov, T. L., Horton, M. W. & Bergelson, J. Genomic variability as a driver of plant-
600 pathogen coevolution? *Current Opinion in Plant Biology* **18**, 24-30,
601 doi:10.1016/j.pbi.2013.12.003 (2014).

602

603 5 Apanius, V., Penn, D., Slev, P. R., Ruff, L. R. & Potts, W. K. The Nature of Selection on
604 the Major Histocompatibility Complex. *Crit Rev Immunol* **37**, 75-120,
605 doi:10.1615/CritRevImmunol.v37.i2-6.10 (2017).

606

607 6 Lenz, T. L., Hafer, N., Samonte, I. E., Yeates, S. E. & Milinski, M. Cryptic haplotype-
608 specific gamete selection yields offspring with optimal MHC immune genes. *Evolution*
609 **72**, 2478-2490, doi:10.1111/evo.13591 (2018).

610

611 7 Penman, B. S. & Gupta, S. Detecting signatures of past pathogen selection on human
612 HLA loci: are there needles in the haystack? *Parasitology* **145**, 731-739,
613 doi:10.1017/S0031182017001159 (2018).

614

615 8 Koenig, D. *et al.* Long-term balancing selection drives evolution of immunity genes in
616 *Capsella*. *eLife* **8**, doi:10.7554/eLife.43606 (2019).

617

618 9 Guoy, A. & Excoffier, L. Polygenic Patterns of Adaptive Introgression in Modern
619 Humans Are Mainly Shaped by Response to Pathogens. *Mol. Biol. Evol.* **37**, 1420-1433,
620 doi:10.1093/molbev/msz306 (2020).

621

622 10 Janzen, D. H. When Is It Coevolution. *Evolution* **34**, 611-612, doi:Doi 10.2307/2408229
623 (1980).

624

625 11 Woolhouse, M. E. J., Webster, J. P., Domingo, E., Charlesworth, B. & Levin, B. R.
626 Biological and biomedical implications of the co-evolution of pathogens and their
627 hosts. *Nature Genetics* **32**, 569-577 (2002).

628

629 12 Kiestler, A. R., Lande, R. & Schemske, D. W. Models of coevolution and speciation in
630 plants and their pollinators. *Am. Nat.* **124**, 220-243 (1984).

631

632 13 Wade, M. J. The co-evolutionary genetics of ecological communities. *Nat. Rev. Genet.*
633 **8**, 185-195 (2007).

634

- 635 14 de Vienne, D. M. *et al.* Cospeciation vs host-shift speciation: methods for testing,
636 evidence from natural associations and relation to coevolution. *New Phytologist* **198**,
637 347-385, doi:10.1111/nph.12150 (2013).
638
- 639 15 Ebert, D. Host-parasite coevolution: Insights from the Daphnia-parasite model system.
640 *Curr. Opin. Microbiol.* **11**, 290-301 (2008).
641
- 642 16 Charlesworth, D. Balancing selection and its effects on sequences in nearby genome
643 regions. *PLoS Genet.* **2**, 379-384, doi:10.1371/journal.pgen.0020064 (2006).
644 **This is an authoritative review on the population genetics of balancing selection.**
645
- 646 17 Zivkovic, D., John, S., Verin, M., Stephan, W. & Tellier, A. Neutral genomic signatures of
647 host-parasite coevolution. *BMC Evol Biol* **19**, doi:10.1186/s12862-019-1556-3 (2019).
648
- 649 18 Maynard Smith, J. & Szathmáry, E. *The Major Transitions in Evolution.* (Oxford
650 University Press, 1995).
651
- 652 19 Dodds, P. N. & Rathjen, J. P. Plant immunity: towards an integrated view of plant-
653 pathogen interactions. *Nat. Rev. Genet.* **11**, 539-548, doi:10.1038/nrg2812 (2010).
654
- 655 20 Sironi, M., Cagliani, R., Forni, D. & Clerici, M. Evolutionary insights into host-pathogen
656 interactions from mammalian sequence data. *Nat. Rev. Genet.* **16**, 224-236,
657 doi:10.1038/nrg3905 (2015).
658
- 659 21 Tellier, A., Moreno-Gamez, S. & Stephan, W. Speed of adaptation and genomic
660 footprints of host-parasite coevolution under arms race and trench warfare dynamics.
661 *Evolution* **68**, 2211-2224, doi:10.1111/evo.12427 (2014).
662
- 663 22 Thrall, P. H., Barrett, L. G., Dodds, P. N. & Burdon, J. J. Epidemiological and
664 Evolutionary Outcomes in Gene-for-Gene and Matching Allele Models. *Frontiers in*
665 *Plant Science* **6**, doi:10.3389/fpls.2015.01084 (2016).
666
- 667 23 Ebert, D. Open questions: what are the genes underlying antagonistic coevolution?
668 *Bmc Biology* **16**, doi:10.1186/s12915-018-0583-7 (2018).
669
- 670 24 Fischer, M. C., Foll, M., Heckel, G. & Excoffier, L. Continental-Scale Footprint of
671 Balancing and Positive Selection in a Small Rodent (*Microtus arvalis*). *Plos One* **9**,
672 doi:10.1371/journal.pone.0112332 (2014).
673
- 674 25 Leffler, E. M. *et al.* Multiple Instances of Ancient Balancing Selection Shared Between
675 Humans and Chimpanzees. *Science* **339**, 1578-1582, doi:10.1126/science.1234070
676 (2013).
677 **This study reports on many sites with TSPs and ancient balancing selection in genomes**
678 **of humans and chimpanzee.**
679
- 680 26 Key, F. M., Teixeira, J. C., de Filippo, C. & Andres, A. M. Advantageous diversity
681 maintained by balancing selection in humans. *Curr Opin Genet Dev* **29**, 45-51,
682 doi:10.1016/j.gde.2014.08.001 (2014).
683

- 684 27 Schweizer, R. M. *et al.* Natural Selection and Origin of a Melanistic Allele in North
685 American Gray Wolves. *Mol. Biol. Evol.* **35**, 1190-1209, doi:10.1093/molbev/msy031
686 (2018).
687
- 688 28 Bitarello, B. D. *et al.* Signatures of Long-Term Balancing Selection in Human Genomes.
689 *Genome Biology and Evolution* **10**, 939-955, doi:10.1093/gbe/evy054 (2018).
690
- 691 29 Cheng, X. H. & DeGiorgio, M. Detection of Shared Balancing Selection in the Absence
692 of Trans-Species Polymorphism. *Mol. Biol. Evol.* **36**, 177-199,
693 doi:10.1093/molbev/msy202 (2019).
694
- 695 30 Enard, D., Cai, L., Gwennap, C. & Petrov, D. A. Viruses are a dominant driver of protein
696 adaptation in mammals. *Elife* **5**, doi:10.7554/eLife.12469 (2016).
697
- 698 31 Schirrmann, M. K. *et al.* Genomewide signatures of selection in *Epichloe* reveal
699 candidate genes for host specialization. *Mol. Ecol.* **27**, 3070-3086,
700 doi:10.1111/mec.14585 (2018).
701
- 702 32 Persoons, A. *et al.* The escalatory Red Queen: Population extinction and replacement
703 following arms race dynamics in poplar rust. *Mol. Ecol.* **26**, 1902-1918,
704 doi:10.1111/mec.13980 (2017).
705
- 706 33 Mohd-Assaad, N., McDonald, B. A. & Croll, D. Genome-Wide Detection of Genes Under
707 Positive Selection in Worldwide Populations of the Barley Scald Pathogen. *Genome*
708 *Biology and Evolution* **10**, 1315-1332, doi:10.1093/gbe/evy087 (2018).
709
- 710 34 Badouin, H. *et al.* Widespread selective sweeps throughout the genome of model
711 plant pathogenic fungi and identification of effector candidates. *Mol. Ecol.* **26**, 2041-
712 2062, doi:10.1111/mec.13976 (2017).
713
- 714 35 Obbard, D. J., Gordon, K. H. J., Buck, A. H. & Jiggins, F. M. The evolution of RNAi as a
715 defence against viruses and transposable elements. *Phil. Trans. R. Soc. B* **364**, 99-115,
716 doi:10.1098/Rstb.2008.0168 (2009).
717
- 718 36 Barrick, J. E. & Lenski, R. E. Genome dynamics during experimental evolution. *Nat. Rev.*
719 *Genet.* **14**, 827-839, doi:10.1038/nrg3564 (2013).
720
- 721 37 Hermisson, J. & Pennings, P. S. Soft sweeps and beyond: understanding the patterns
722 and probabilities of selection footprints under rapid adaptation. *Methods Ecol Evol* **8**,
723 700-716, doi:10.1111/2041-210x.12808 (2017).
724
- 725 38 Hahn, M. W. *Molecular Population Genetics*. (Sinauer Associates, 2018).
726
- 727 39 McCarthy, K. R., Kirmaier, A., Autissier, P. & Johnson, W. E. Evolutionary and
728 Functional Analysis of Old World Primate TRIM5 Reveals the Ancient Emergence of
729 Primate Lentiviruses and Convergent Evolution Targeting a Conserved Capsid
730 Interface. *PLoS Pathog.* **11**, doi:10.1371/journal.ppat.1005085 (2015).
731
- 732 40 Elena, S. F., Cooper, V. S. & Lenski, R. E. Punctuated evolution caused by selection of
733 rare beneficial mutations. *Science* **272**, 1802-1804 (1996).

734
735 41 Anderson, T. J. C. *et al.* Population Parameters Underlying an Ongoing Soft Sweep in
736 Southeast Asian Malaria Parasites. *Mol. Biol. Evol.* **34**, 131-144,
737 doi:10.1093/molbev/msw228 (2017).
738

739 42 Garud, N. R., Messer, P. W., Buzbas, E. O. & Petrov, D. A. Recent Selective Sweeps in
740 North American *Drosophila melanogaster* Show Signatures of Soft Sweeps. *PLoS*
741 *Genet.* **11**, e1005004, doi:10.1371/journal.pgen.1005004 (2015).
742

743 43 Sanchez-Vallet, A. *et al.* The Genome Biology of Effector Gene Evolution in
744 Filamentous Plant Pathogens. *Ann. Rev. Phytopathol.* **56**, 21-40, doi:10.1146/annurev-
745 phyto-080516-035303 (2018).
746

747 44 Moxon, E. R., Rainey, P. B., Nowak, M. A. & Lenski, R. E. Adaptive evolution of highly
748 mutable loci in pathogenic bacteria. *Curr. Biol.* **4**, 24-33 (1994).
749

750 45 Raffaele, S. *et al.* Genome Evolution Following Host Jumps in the Irish Potato Famine
751 Pathogen Lineage. *Science* **330**, 1540-1543, doi:10.1126/science.1193070 (2010).
752

753 46 Croll, D. & McDonald, B. A. The Accessory Genome as a Cradle for Adaptive Evolution
754 in Pathogens. *PLoS Pathog.* **8**, doi:10.1371/journal.ppat.1002608 (2012).
755

756 47 Clarke, B. C. The ecological relationship of host-parasite relationships. in *Genetic*
757 *aspects of host-parasite relationships* (eds A.E.R. Taylor & R.M. Muller) 87-104
758 (Blackwell, Oxford, 1976).
759

760 48 Hamilton, W. D., Axelrod, R. & Tanese, R. Sexual reproduction as an adaptation to
761 resist parasites. *Proc. Nat. Acad. Sci. USA* **87**, 3566-3573 (1990).
762

763 49 Fenton, A., Antonovics, J. & Brockhurst, Michael A. Inverse-Gene-for-Gene Infection
764 Genetics and Coevolutionary Dynamics. *Am. Nat.* **174**, E230-E242,
765 doi:10.1086/645087 (2009).
766

767 50 Schmid-Hempel, P. *Evolutionary parasitology: the integrated study of infections,*
768 *immunology, ecology, and genetics.* (Oxford University Press, 2011).
769 **This is a comprehensive text book summarizing the field of host-parasite evolution and**
770 **coevolution.**
771

772 51 Ben Khalifa, M., Simon, V., Fakhfakh, H. & Moury, B. Tunisian Potato virus Y isolates
773 with unnecessary pathogenicity towards pepper: support for the matching allele
774 model in eIF4E resistance-potyvirus interactions. *Plant Pathol.* **61**, 441-447,
775 doi:10.1111/j.1365-3059.2011.02540.x (2012).
776

777 52 Luijckx, P., Fienberg, H., Duneau, D. & Ebert, D. A Matching-Allele Model Explains Host
778 Resistance to Parasites. *Curr. Biol.* **23**, 1085-1088, doi:10.1016/J.Cub.2013.04.064
779 (2013).
780 **This study provides an early example of a well worked out matching-allele host-**
781 **parasite interaction matrix.**
782

783 53 Bento, G. *et al.* The genetic basis of resistance and matching-allele interactions of a
784 host-parasite system: The *Daphnia magna*-*Pasteuria ramosa* model. *PLoS Genet.* **13**,
785 doi:10.1371/journal.pgen.1006596 (2017).
786

787 54 King, K. C., Jokela, J. & Lively, C. M. Parasites, Sex, and Clonal Diversity in Natural Snail
788 Populations. *Evolution* **65**, 1474-1481, doi:Doi 10.1111/J.1558-5646.2010.01215.X
789 (2011).
790

791 55 Ashby, B. & Boots, M. Multi-mode fluctuating selection in host-parasite coevolution.
792 *Ecol. Lett.* **20**, 357-365, doi:10.1111/ele.12734 (2017).
793

794 56 Papkou, A. *et al.* The genomic basis of Red Queen dynamics during rapid reciprocal
795 host-pathogen coevolution. *Proc. Natl. Acad. Sci. USA* **116**, 923-928,
796 doi:10.1073/pnas.1810402116 (2019).
797 **This study of experimental evolution with nematodes and a bacterial pathogen**
798 **demonstrate the complexity of coevolutionary interactions emerging in seemingly**
799 **simple systems.**
800

801 57 Koskella, B. & Lively, C. M. Evidence for Negative Frequency-Dependent Selection
802 during Experimental Coevolution of a Freshwater Snail and a Sterilizing Trematode.
803 *Evolution* **63**, 2213-2221, doi:10.1111/J.1558-5646.2009.00711.X (2009).
804

805 58 Lively, C. M. Habitat Heterogeneity, Host Population Structure, and Parasite Local
806 Adaptation. *J. Hered.* **109**, 29-37, doi:10.1093/jhered/esx100 (2018).
807

808 59 Ejsmond, M. J., Babik, W. & Radwan, J. MHC allele frequency distributions under
809 parasite-driven selection: A simulation model. *BMC Evol Biol* **10**, doi:10.1186/1471-
810 2148-10-332 (2010).
811

812 60 Fijarczyk, A. & Babik, W. Detecting balancing selection in genomes: limits and
813 prospects. *Mol. Ecol.* **24**, 3529-3545, doi:10.1111/mec.13226 (2015).
814

815 61 Bubb, K. L. *et al.* Scan of human genome reveals no new loci under ancient balancing
816 selection. *Genetics* **173**, 2165-2177, doi:10.1534/genetics.106.055715 (2006).
817

818 62 Cagliani, R. *et al.* The signature of long-standing balancing selection at the human
819 defensin beta-1 promoter. *Genome Biol* **9**, R143, doi:10.1186/gb-2008-9-9-r143
820 (2008).
821

822 63 Fumagalli, M. *et al.* Widespread balancing selection and pathogen-driven selection at
823 blood group antigen genes. *Genome Research* **19**, 199-212,
824 doi:10.1101/Gr.082768.108 (2009).
825

826 64 Segurel, L. *et al.* The ABO blood group is a trans-species polymorphism in primates.
827 *Proc. Natl. Acad. Sci. USA* **109**, 18493-18498, doi:10.1073/pnas.1210603109 (2012).
828

829 65 Bergelson, J., Kreitman, M., Stahl, E. A. & Tian, D. C. Evolutionary dynamics of plant R-
830 genes. *Science* **292**, 2281-2285, doi:10.1126/science.1061337 (2001).
831

- 832 66 Hoerger, A. C. *et al.* Balancing Selection at the Tomato RCR3 Guardee Gene Family
833 Maintains Variation in Strength of Pathogen Defense. *PLoS Genet.* **8**,
834 doi:10.1371/journal.pgen.1002813 (2012).
835
- 836 67 Llaurens, V., Whibley, A. & Joron, M. Genetic architecture and balancing selection: the
837 life and death of differentiated variants. *Mol. Ecol.* **26**, 2430-2448,
838 doi:10.1111/mec.14051 (2017).
839
- 840 68 Croze, M. *et al.* A genome-wide scan for genes under balancing selection in *Drosophila*
841 *melanogaster*. *BMC Evol Biol* **17**, doi:10.1186/s12862-016-0857-z (2017).
842
- 843 69 Buckley, J., Holub, E. B., Koch, M. A., Vergeer, P. & Mable, B. K. Restriction associated
844 DNA-genotyping at multiple spatial scales in *Arabidopsis lyrata* reveals signatures of
845 pathogen-mediated selection. *Bmc Genomics* **19**, doi:10.1186/s12864-018-4806-7
846 (2018).
847
- 848 70 Wu, Q. *et al.* Long-term balancing selection contributes to adaptation in *Arabidopsis*
849 and its relatives. *Genome Biology* **18**, doi:10.1186/s13059-017-1342-8 (2017).
850
- 851 71 Unckless, R. L., Howick, V. M. & Lazzaro, B. P. Convergent Balancing Selection on an
852 Antimicrobial Peptide in *Drosophila*. *Curr. Biol.* **26**, 257-262,
853 doi:10.1016/j.cub.2015.11.063 (2016).
854
- 855 72 Bartoli, C. & Roux, F. Genome-Wide Association Studies In Plant Pathosystems: Toward
856 an Ecological Genomics Approach. *Frontiers in Plant Science* **8**,
857 doi:10.3389/fpls.2017.00763 (2017).
858
- 859 73 Power, R. A., Parkhill, J. & de Oliveira, T. Microbial genome-wide association studies:
860 lessons from human GWAS. *Nat. Rev. Genet.* **18**, 41-50, doi:10.1038/nrg.2016.132
861 (2017).
862
- 863 74 Thomas, J. C., Godfrey, P. A., Feldgarden, M. & Robinson, A. Candidate Targets of
864 Balancing Selection in the Genome of *Staphylococcus aureus*. *Mol. Biol. Evol.* **29**, 1175-
865 1186, doi:10.1093/molbev/msr286 (2012).
866
- 867 75 Zhang, L. F., Thomas, J. C., Didelot, X. & Robinson, D. A. Molecular Signatures Identify a
868 Candidate Target of Balancing Selection in an *arcD*-Like Gene of *Staphylococcus*
869 *epidermidis*. *J. Mol. Evol.* **75**, 43-54, doi:10.1007/s00239-012-9520-5 (2012).
870
- 871 76 Guttman, D. S., Gropp, S. J., Morgan, R. L. & Wang, P. W. Diversifying selection drives
872 the evolution of the type III secretion system pilus of *Pseudomonas syringae*. *Mol. Biol.*
873 *Evol.* **23**, 2342-2354, doi:10.1093/molbev/msl103 (2006).
874
- 875 77 Castillo, J. A. & Agathos, S. N. A genome-wide scan for genes under balancing selection
876 in the plant pathogen *Ralstonia solanacearum*. *BMC Evol Biol* **19**, doi:10.1186/s12862-
877 019-1456-6 (2019).
878
- 879 78 Ryabov, E. V. *et al.* Dynamic evolution in the key honey bee pathogen deformed wing
880 virus: Novel insights into virulence and competition using reverse genetics. *PLoS Biol.*
881 **17**, doi:10.1371/journal.pbio.3000502 (2019).

882
883 79 Andras, J. P., Fields, P. D., Du Pasquier, L., Fredericksen, M. & Ebert, D. Genome-wide
884 association analysis identifies a genetic basis of infectivity in a model bacterial
885 pathogen. . *Mol. Biol. Evol.* **(in press)**, doi:10.1093/molbev/msaa173 (2020).
886
887 80 Brackney, D. E., Beane, J. E. & Ebel, G. D. RNAi Targeting of West Nile Virus in
888 Mosquito Midguts Promotes Virus Diversification. *PLoS Pathog.* **5**,
889 doi:10.1371/journal.ppat.1000502 (2009).
890
891 81 Brackney, D. E., Schirtzinger, E. E., Harrison, T. D., Ebel, G. D. & Hanley, K. A.
892 Modulation of Flavivirus Population Diversity by RNA Interference. *J. Virol.* **89**, 4035-
893 4039, doi:10.1128/Jvi.02612-14 (2015).
894
895 82 Obbard, D. J., Jiggins, F. M., Halligan, D. L. & Little, T. J. Natural selection drives
896 extremely rapid evolution in antiviral RNAi genes. *Curr. Biol.* **16**, 580-585, doi:Doi
897 10.1016/J.Cub.2006.01.065 (2006).
898
899 83 Araki, H. *et al.* Presence/absence polymorphism for alternative pathogenicity islands in
900 *Pseudomonas viridiflava*, a pathogen of *Arabidopsis*. *Proc. Natl. Acad. Sci. USA* **103**,
901 5887-5892, doi:10.1073/pnas.0601431103 (2006).
902
903 84 Nygaard, S. *et al.* Long- and Short-Term Selective Forces on Malaria Parasite Genomes.
904 *PLoS Genet.* **6**, doi:10.1371/journal.pgen.1001099 (2010).
905
906 85 Ochola, L. I. *et al.* Allele Frequency-Based and Polymorphism-Versus-Divergence
907 Indices of Balancing Selection in a New Filtered Set of Polymorphic Genes in
908 *Plasmodium falciparum*. *Mol. Biol. Evol.* **27**, 2344-2351, doi:10.1093/molbev/msq119
909 (2010).
910
911 86 Salathe, M., Kouyos, R. D., Regoes, R. R. & Bonhoeffer, S. Rapid parasite adaptation
912 drives selection for high recombination rates. *Evolution* **62**, 295-300 (2008).
913
914 87 Gokhale, C. S., Papkou, A., Traulsen, A. & Schulenburg, H. Lotka-Volterra dynamics kills
915 the Red Queen: population size fluctuations and associated stochasticity dramatically
916 change host-parasite coevolution. *BMC Evol Biol* **13**, doi:10.1186/1471-2148-13-254
917 (2013).
918
919 88 Schenk, H., Schulenburg, H. & Traulsen, A. How long do Red Queen dynamics survive
920 under genetic drift? A comparative analysis of evolutionary and eco-evolutionary
921 models. *BMC Evol Biol* **20**, doi:10.1186/s12862-019-1562-5 (2020).
922
923 89 MacPherson, A., Keeling, M. J. & Otto, S. P. Coevolution does not slow the rate of loss
924 of heterozygosity in a stochastic host-parasite model with constant population size.
925 *bioRxiv*, doi:10.1101/2020.04.07.024661 (2020).
926
927 90 Thrall, P. H. & Burdon, J. J. Effect of resistance variation in a natural plant host-
928 pathogen metapopulation on disease dynamics. *Plant Pathol.* **49**, 767-773, doi:DOI
929 10.1046/j.1365-3059.2000.00523.x (2000).
930

931 91 Brown, J. K. M. & Tellier, A. Plant-Parasite Coevolution: Bridging the Gap between
932 Genetics and Ecology. *Annu Rev Phytopathol* **49**, 345-367, doi:10.1146/Annurev-
933 Phyto-072910-095301 (2011).
934

935 92 Radwan, J., Babik, W., Kaufman, J., Lenz, T. L. & Winternitz, J. Advances in the
936 Evolutionary Understanding of MHC Polymorphism. *Trends Genetics* **36**, 298-311,
937 doi:10.1016/j.tig.2020.01.008 (2020).
938

939 93 Charlesworth, B., Nordborg, M. & Charlesworth, D. The effects of local selection,
940 balanced polymorphism and background selection on equilibrium patterns of genetic
941 diversity in subdivided populations. *Genet. Res. Camb.* **70**, 155-174 (1997).
942

943 94 Eizaguirre, C., Lenz, T. L., Kalbe, M. & Milinski, M. Divergent selection on locally
944 adapted major histocompatibility complex immune genes experimentally proven in
945 the field. *Ecol. Lett.* **15**, 723-731, doi:10.1111/j.1461-0248.2012.01791.x (2012).
946

947 95 Rico, Y. *et al.* Spatial patterns of immunogenetic and neutral variation underscore the
948 conservation value of small, isolated American badger populations. *Evolutionary*
949 *Applications* **9**, 1271-1284, doi:10.1111/eva.12410 (2016).
950

951 96 Jousimo, J. *et al.* Ecological and evolutionary effects of fragmentation on infectious
952 disease dynamics. *Science* **344**, 1289-1293, doi:10.1126/science.1253621 (2014).
953

954 97 Crispo, E. *et al.* The evolution of the major histocompatibility complex in upstream
955 versus downstream river populations of the longnose dace. *Ecol Evol* **7**, 3297-3311,
956 doi:10.1002/ece3.2839 (2017).
957

958 98 Keller, M. F. *et al.* Trans-ethnic meta-analysis of white blood cell phenotypes. *Human*
959 *Molecular Genetics* **23**, 6944-6960, doi:10.1093/hmg/ddu401 (2014).
960

961 99 Morgan, A. D., Gandon, S. & Buckling, A. The effect of migration on local adaptation in
962 a coevolving host-parasite system. *Nature* **437**, 253-256 (2005).
963

964 100 Thrall, P. H. *et al.* Rapid genetic change underpins antagonistic coevolution in a natural
965 host-pathogen metapopulation. *Ecol. Lett.* **15**, 425-435, doi:10.1111/J.1461-
966 0248.2012.01749.X (2012).
967

968 101 Kawecki, T. J. & Ebert, D. Conceptual issues in local adaptation. *Ecol. Lett.* **7**, 1225-1241
969 (2004).
970

971 102 Croll, D. & McDonald, B. A. The genetic basis of local adaptation for pathogenic fungi
972 in agricultural ecosystems. *Mol. Ecol.* **26**, 2027-2040, doi:10.1111/mec.13870 (2017).
973

974 103 Laine, A. L., Burdon, J. J., Dodds, P. N. & Thrall, P. H. Spatial variation in disease
975 resistance: from molecules to metapopulations. *J. Ecol.* **99**, 96-112,
976 doi:10.1111/J.1365-2745.2010.01738.X (2011).
977

978 104 Bolnick, D. I. & Stutz, W. E. Frequency dependence limits divergent evolution by
979 favouring rare immigrants over residents. *Nature* **546**, 285-288,
980 doi:10.1038/nature22351 (2017).

981 **This experimental study with fish shows that rare immigrants have an advantage over**
982 **resident genotypes. The study demonstrates elegantly that resistance genes have**
983 **higher effective migration rates.**
984

985 105 Phillips, K. P. *et al.* Immunogenetic novelty confers a selective advantage in host-
986 pathogen coevolution. *Proc. Natl. Acad. Sci. USA* **115**, 1552-1557,
987 doi:10.1073/pnas.1708597115 (2018).
988

989 106 Rico, Y., Morris-Pocock, J., Zigouris, J., Nocera, J. J. & Kyle, C. J. Lack of Spatial
990 Immunogenetic Structure among Wolverine (*Gulo gulo*) Populations Suggestive of
991 Broad Scale Balancing Selection. *Plos One* **10**, doi:10.1371/journal.pone.0140170
992 (2015).
993

994 107 Leducq, J. B. *et al.* Effect of balancing selection on spatial genetic structure within
995 populations: theoretical investigations on the self-incompatibility locus and empirical
996 studies in *Arabidopsis halleri*. *Heredity* **106**, 319-329, doi:10.1038/hdy.2010.68 (2011).
997

998 108 Castric, V., Bechsgaard, J., Schierup, M. H. & Vekemans, X. Repeated Adaptive
999 Introgression at a Gene under Multiallelic Balancing Selection. *PLoS Genet.* **4**,
1000 doi:10.1371/journal.pgen.1000168 (2008).
1001

1002 109 Hoban, S. *et al.* Finding the Genomic Basis of Local Adaptation: Pitfalls, Practical
1003 Solutions, and Future Directions. *Am. Nat.* **188**, 379-397, doi:10.1086/688018 (2016).
1004

1005 110 Borg, A. A., Pedersen, S. A., Jensen, H. & Westerdahl, H. Variation in MHC genotypes in
1006 two populations of house sparrow (*Passer domesticus*) with different population
1007 histories. *Ecol Evol* **1**, 145-159, doi:10.1002/ece3.13 (2011).
1008

1009 111 Novembre, J. *et al.* Genes mirror geography within Europe. *Nature* **456**, 98-U95,
1010 doi:10.1038/nature07331 (2008).
1011

1012 112 Fields, P. D., Reisser, C., Dukic, M., Haag, C. R. & Ebert, D. Genes mirror geography in
1013 *Daphnia magna*. *Mol. Ecol.* **24**, 4521-4536, doi:10.1111/mec.13324 (2015).
1014

1015 113 Thompson, J. N. *The Geographic Mosaic of Coevolution*. (University of Chicago Press,
1016 2005).
1017

1018 114 Laine, A. L., Barres, B., Numminen, E. & Siren, J. P. Variable opportunities for
1019 outcrossing result in hotspots of novel genetic variation in a pathogen
1020 metapopulation. *eLife* **8**, doi:10.7554/eLife.47091 (2019).
1021

1022 115 Klein, J. *Immunology*. (Blackwell, 1990).
1023

1024 116 Lenz, T. L., Eizaguirre, C., Kalbe, M. & Milinski, M. Evaluating Patterns of Convergent
1025 Evolution and Trans-Species Polymorphism at Mhc Immunogenes in Two Sympatric
1026 Stickleback Species. *Evolution* **67**, 2400-2412, doi:10.1111/evo.12124 (2013).
1027 **This study demonstrates TSP in two sympatric stickleback fish sharing the same**
1028 **parasites. The authors were able to rule out convergent evolution as an alternative**
1029 **explanation for TSP.**
1030

1031 117 Tesicky, M. & Vinkler, M. Trans-Species Polymorphism in Immune Genes: General
1032 Pattern or MHC-Restricted Phenomenon? *J Immunol Res*, doi:10.1155/2015/838035
1033 (2015).
1034

1035 118 Mboup, M., Fischer, I., Lainer, H. & Stephan, W. Trans-Species Polymorphism and
1036 Allele-Specific Expression in the CBF Gene Family of Wild Tomatoes. *Mol. Biol. Evol.* **29**,
1037 3641-3652, doi:10.1093/molbev/mss176 (2012).
1038

1039 119 Novikova, P. Y. *et al.* Sequencing of the genus *Arabidopsis* identifies a complex history
1040 of nonbifurcating speciation and abundant trans-specific polymorphism. *Nature*
1041 *Genetics* **48**, 1077–1082, doi:10.1038/ng.3617 (2016).
1042

1043 120 Azevedo, L., Serrano, C., Amorim, A. & Cooper, D. N. Trans-species polymorphism in
1044 humans and the great apes is generally maintained by balancing selection that
1045 modulates the host immune response. *Hum Genomics* **9**, doi:10.1186/s40246-015-
1046 0043-1 (2015).
1047

1048 121 Lenz, T. L. Computational Prediction of Mhc li-Antigen Binding Supports Divergent
1049 Allele Advantage and Explains Trans-Species Polymorphism. *Evolution* **65**, 2380-2390,
1050 doi:10.1111/j.1558-5646.2011.01288.x (2011).
1051

1052 122 Eizaguirre, C., Lenz, T. L., Kalbe, M. & Milinski, M. Rapid and adaptive evolution of MHC
1053 genes under parasite selection in experimental vertebrate populations. *Nature*
1054 *Communications* **3**, doi:10.1038/ncomms1632 (2012).
1055

1056 123 Gao, Z. Y., Przeworski, M. & Sella, G. Footprints of ancient-balanced polymorphisms in
1057 genetic variation data from closely related species. *Evolution* **69**, 431-446,
1058 doi:10.1111/evo.12567 (2015).
1059

1060 124 Hedrick, P. W. Pathogen resistance and genetic variation at MHC loci. *Evolution* **56**,
1061 1902-1908 (2002).
1062

1063 125 Eizaguirre, C. & Lenz, T. L. Major histocompatibility complex polymorphism: dynamics
1064 and consequences of parasite-mediated local adaptation in fishes. *J. Fish Biol.* **77**,
1065 2023-2047, doi:10.1111/j.1095-8649.2010.02819.x (2010).
1066

1067 126 Osborne, M. J., Pilger, T. J., Lusk, J. D. & Turner, T. F. Spatio-temporal variation in
1068 parasite communities maintains diversity at the major histocompatibility complex class
1069 II in the endangered Rio Grande silvery minnow. *Mol. Ecol.* **26**, 471-489,
1070 doi:10.1111/mec.13936 (2017).
1071

1072 127 Daugherty, M. D. & Malik, H. S. Rules of Engagement: Molecular Insights from Host-
1073 Virus Arms Races. *Annu Rev Genet* **46**, 677-700, doi:10.1146/annurev-genet-110711-
1074 155522 (2012).
1075

1076 128 Cagliani, R. *et al.* A Positively Selected Apobec3h Haplotype Is Associated with Natural
1077 Resistance to Hiv-1 Infection. *Evolution* **65**, 3311-3322, doi:10.1111/j.1558-
1078 5646.2011.01368.x (2011).
1079

1080 129 Davis, Z. H. *et al.* Global Mapping of Herpesvirus-Host Protein Complexes Reveals a
1081 Transcription Strategy for Late Genes. *Molecular Cell* **57**, 349-360,
1082 doi:10.1016/j.molcel.2014.11.026 (2015).
1083

1084 130 Lozano-Torres, J. L. *et al.* Dual disease resistance mediated by the immune receptor
1085 Cf-2 in tomato requires a common virulence target of a fungus and a nematode. *Proc.*
1086 *Natl. Acad. Sci. USA* **109**, 10119-10124, doi:10.1073/pnas.1202867109 (2012).
1087

1088 131 Wessling, R. *et al.* Convergent Targeting of a Common Host Protein-Network by
1089 Pathogen Effectors from Three Kingdoms of Life. *Cell Host & Microbe* **16**, 364-375,
1090 doi:10.1016/j.chom.2014.08.004 (2014).
1091

1092 132 Wegner, K. M., Kalbe, M., Kurtz, J., Reusch, T. B. H. & Milinski, M. Parasite selection for
1093 immunogenetic optimality. *Science* **301**, 1343-1343 (2003).
1094

1095 133 Matzaraki, V., Kumar, V., Wijmenga, C. & Zhernakova, A. The MHC locus and genetic
1096 susceptibility to autoimmune and infectious diseases. *Genome Biology* **18**,
1097 doi:10.1186/s13059-017-1207-1 (2017).
1098

1099 134 Karasov, T. L., Barrett, L., Hershberg, R. & Bergelson, J. Similar levels of gene content
1100 variation observed for *Pseudomonas syringae* populations extracted from single and
1101 multiple host species. *Plos One* **12**, doi:10.1371/journal.pone.0184195 (2017).
1102

1103 135 Bechsgaard, J., Jorgensen, T. H. & Schierup, M. H. Evidence for Adaptive Introgression
1104 of Disease Resistance Genes Among Closely Related Arabidopsis Species. *G3-Genes*
1105 *Genomes Genetics* **7**, 2677-2683, doi:10.1534/g3.117.043984 (2017).
1106

1107 136 Gluck-Thaler, E. & Slot, J. C. Dimensions of Horizontal Gene Transfer in Eukaryotic
1108 Microbial Pathogens. *PLoS Pathog.* **11**, doi:ARTN e1005156
1109 10.1371/journal.ppat.1005156 (2015).
1110

1111 137 Campbell, M. C., Ashong, B., Teng, S. L., Harvey, J. & Cross, C. N. Multiple selective
1112 sweeps of ancient polymorphisms in and around LT alpha located in the MHC class III
1113 region on chromosome 6. *BMC Evol Biol* **19**, 218, doi:10.1186/s12862-019-1557-2
1114 (2019).
1115

1116 138 Karasov, T. L. *et al.* The long-term maintenance of a resistance polymorphism through
1117 diffuse interactions. *Nature* **512**, 436-U472, doi:10.1038/nature13439 (2014).
1118

1119 139 Rabajante, J. F. *et al.* Red Queen dynamics in multi-host and multi-parasite interaction
1120 system. *Scientific Reports* **5**, doi:10.1038/srep10004 (2015).
1121

1122 140 Kamath, P. L., Turner, W. C., Kusters, M. & Getz, W. M. Parasite-mediated selection
1123 drives an immunogenetic trade-off in plains zebras (*Equus quagga*). *Proc. R. Soc. B* **281**,
1124 doi:10.1098/rspb.2014.0077 (2014).
1125

1126 141 Nadeem, A. & Wahl, L. M. Prophage as a genetic reservoir: Promoting diversity and
1127 driving innovation in the host community. *Evolution* **71**, 2080-2089,
1128 doi:10.1111/evo.13287 (2017).
1129

- 1130 142 Fortuna, M. A. *et al.* Coevolutionary dynamics shape the structure of bacteria-phage
1131 infection networks. *Evolution* **73**, 1001-1011, doi:10.1111/evo.13731 (2019).
1132
- 1133 143 Silva, J. C. *et al.* Genome sequences reveal divergence times of malaria parasite
1134 lineages. *Parasitology* **138**, 1737-1749, doi:10.1017/S0031182010001575 (2011).
1135
- 1136 144 Galen, S. C. *et al.* The polyphyly of Plasmodium: comprehensive phylogenetic analyses
1137 of the malaria parasites (order Haemosporida) reveal widespread taxonomic conflict.
1138 *Roy Soc Open Sci* **5**, doi:10.1098/rsos.171780 (2018).
1139
- 1140 145 Otto, T. D. *et al.* Genomes of all known members of a Plasmodium subgenus reveal
1141 paths to virulent human malaria. *Nat Microbiol* **3**, 687-697, doi:10.1038/s41564-018-
1142 0162-2 (2018).
1143
- 1144 146 Pacheco, M. A. *et al.* Mode and Rate of Evolution of Haemosporidian Mitochondrial
1145 Genomes: Timing the Radiation of Avian Parasites. *Mol. Biol. Evol.* **35**, 383-403,
1146 doi:10.1093/molbev/msx285 (2018).
1147
- 1148 147 Bartha, I. *et al.* A genome-to-genome analysis of associations between human genetic
1149 variation, HIV-1 sequence diversity, and viral control. *eLife* **2**, doi:10.7554/eLife.01123
1150 (2013).
1151 **This study describes a pioneering method in cogenomics, applied to interacting**
1152 **genomic sites in hosts and parasites.**
1153
- 1154 148 Lees, J. A., Tonkin-Hill, G. & Bentley, S. D. GENOME WATCH Stronger together. *Nat.*
1155 *Rev. Microbiol.* **15**, 516-516, doi:10.1038/nrmicro.2017.95 (2017).
1156
- 1157 149 Märkle, H. & Tellier, A. Inference of coevolutionary dynamics and parameters from
1158 host and parasite polymorphism data of repeated experiments. *PLoS Computational*
1159 *Biology* **16**, e1007668, doi:10.1371/journal.pcbi.1007668 (2020).
1160
- 1161 150 Otto, S. P. & Nuismer, S. L. Species interactions and the evolution of sex. *Science* **304**,
1162 1018-1020 (2004).
1163
- 1164 151 Tellier, A. & Brown, J. K. M. Polymorphism in multilocus host-parasite coevolutionary
1165 interactions. *Genetics* **177**, 1777-1790, doi:10.1534/Genetics.107.074393 (2007).
1166
- 1167 152 Engelstadter, J. & Bonhoeffer, S. Red Queen Dynamics with Non-Standard Fitness
1168 Interactions. *PLoS Computational Biology* **5**, doi:10.1371/Journal.Pcbi.1000469 (2009).
1169
- 1170 153 Best, A. *et al.* The Evolution of Host-Parasite Range. *Am. Nat.* **176**, 63-71,
1171 doi:10.1086/653002 (2010).
1172
- 1173 154 Fenton, A., Antonovics, J. & Brockhurst, M. A. Two-Step Infection Processes Can Lead
1174 to Coevolution between Functionally Independent Infection and Resistance Pathways.
1175 *Evolution* **66**, 2030-2041, doi:10.1111/J.1558-5646.2012.01578.X (2012).
1176
- 1177 155 Kwiatkowski, M., Engelstadter, J. & Vorburger, C. On Genetic Specificity in Symbiont-
1178 Mediated Host-Parasite Coevolution. *Plos Computational Biology* **8**,
1179 doi:10.1371/journal.pcbi.1002633 (2012).

1180
1181 156 Flor, H. H. Host-Parasite Interaction in Flax Rust - Its Genetics and Other Implications.
1182 *Phytopathol.* **45**, 680-685 (1955).
1183
1184 157 Märkle, H., Tellier, A. & John, S. Cross-Species association statistics for genome-wide
1185 studies of host and parasite polymorphism data. *bioRxiv*, 726166, doi:10.1101/726166
1186 (2019).
1187
1188 158 Balmer, O. & Tanner, M. Prevalence and implications of multiple-strain infections.
1189 *Lancet Infect Dis* **11**, 868-878, doi:10.1016/S1473-3099(11)70241-9 (2011).
1190
1191 159 Ansari, M. A. *et al.* Genome-to-genome analysis highlights the effect of the human
1192 innate and adaptive immune systems on the hepatitis C virus. *Nature Genetics* **49**,
1193 666–673, doi:10.1038/ng.3835 (2017).
1194 **This study describes a strong example of the application of cogenomics to find**
1195 **interacting loci in humans infected with hepatitis C-virus.**
1196
1197 160 Lees, J. A. *et al.* Joint sequencing of human and pathogen genomes reveals the
1198 genetics of pneumococcal meningitis. *Nature Communications* **10**,
1199 doi:10.1038/s41467-019-09976-3 (2019).
1200
1201 161 Naret, O. *et al.* Correcting for Population Stratification Reduces False Positive and False
1202 Negative Results in Joint Analyses of Host and Pathogen Genomes. *Frontiers in*
1203 *Genetics* **9**, doi:10.3389/fgene.2018.00266 (2018).
1204
1205 162 Ansari, M. A. *et al.* Interferon lambda 4 impacts the genetic diversity of hepatitis C
1206 virus. *eLife* **8**, doi:10.7554/eLife.42463 (2019).
1207
1208 163 McHenry, M. L. *et al.* Interaction between host genes and Mycobacterium tuberculosis
1209 lineage can affect tuberculosis severity: evidence for coevolution? *PLoS Genet* **16**,
1210 e1008728, doi:10.1371/journal.pgen.1008728 (2020).
1211
1212 164 Wang, M. Y. *et al.* Two-way mixed-effects methods for joint association analysis using
1213 both host and pathogen genomes. *Proc. Natl. Acad. Sci. USA* **115**, E5440-E5449,
1214 doi:10.1073/pnas.1710980115 (2018).
1215 **This study describes the development of a powerful cogenomics method that utilizes**
1216 **data from an interaction matrix of all combinations of host and parasite genotypes to**
1217 **find the genomic sites that underlay the interaction.**
1218
1219 165 Hill, A. V. S., Jepson, A., Plebanski, M. & Gilbert, S. C. Genetic analysis of host-parasite
1220 coevolution in human malaria. *Phil. Trans. R. Soc. B* **352**, 1317-1325 (1997).
1221
1222 166 Heeney, J. L., Dalgleish, A. G. & Weiss, R. A. Origins of HIV and the evolution of
1223 resistance to AIDS. *Science* **313**, 462-466, doi:10.1126/science.1123016 (2006).
1224
1225 167 Hertz, T. *et al.* Mapping the Landscape of Host-Pathogen Coevolution: HLA Class I
1226 Binding and Its Relationship with Evolutionary Conservation in Human and Viral
1227 Proteins. *J. Virol.* **85**, 1310-1321, doi:10.1128/Jvi.01966-10 (2011).
1228

1229 168 Schwander, T., Libbrecht, R. & Keller, L. Supergenes and Complex Phenotypes. *Curr. Biol.* **24**, R288-R294, doi:10.1016/j.cub.2014.01.056 (2014).
1230
1231
1232 169 Lenz, T. L. *et al.* Widespread non-additive and interaction effects within HLA loci
1233 modulate the risk of autoimmune diseases. *Nature Genetics* **47**, 1085–1090,
1234 doi:10.1038/ng.3379 (2015).
1235
1236 170 Salathe, M., Kouyos, R. D. & Bonhoeffer, S. The state of affairs in the kingdom of the
1237 Red Queen. *Trends Ecol. Evol.* **23**, 439-445 (2008).
1238
1239 171 da Silva, J. & Galbraith, J. D. Hill-Robertson interference maintained by Red Queen
1240 dynamics favours the evolution of sex. *J. evol. Biol.* **30**, 994-1010,
1241 doi:10.1111/jeb.13068 (2017).
1242
1243 172 Kubinak, J. L. *et al.* Experimental viral evolution reveals major histocompatibility
1244 complex polymorphisms as the primary host factors controlling pathogen adaptation
1245 and virulence. *Genes Immun* **14**, 365-372, doi:10.1038/gene.2013.27 (2013).
1246
1247 173 Brockhurst, M. A. & Koskella, B. Experimental coevolution of species interactions.
1248 *Trends Ecol. Evol.* **28**, 367-375, doi:10.1016/j.tree.2013.02.009 (2013).
1249
1250 174 Retel, C. *et al.* The feedback between selection and demography shapes genomic
1251 diversity during coevolution. *Sci Adv* **5**, doi:10.1126/sciadv.aax0530 (2019).
1252
1253 175 Figueroa, F., Günther, E. & Klein, J. MHC polymorphism pre-dating speciation. *Nature*
1254 **335**, 265-267, doi:10.1038/335265a0 (1988).
1255
1256 176 Mcdonald, J. H. & Kreitman, M. Adaptive Protein Evolution at the Adh Locus in
1257 *Drosophila*. *Nature* **351**, 652-654, doi:10.1038/351652a0 (1991).
1258
1259 177 Eyre-Walker, A. & Keightley, P. D. Estimating the Rate of Adaptive Molecular Evolution
1260 in the Presence of Slightly Deleterious Mutations and Population Size Change. *Mol.*
1261 *Biol. Evol.* **26**, 2097-2108, doi:10.1093/molbev/msp119 (2009).
1262
1263 178 Cheng, X. & DeGiorgio, M. Detection of Shared Balancing Selection in the Absence of
1264 Trans-Species Polymorphism. *Mol. Biol. Evol.* **36**, 177-199,
1265 doi:10.1093/molbev/msy202 (2018).
1266
1267 179 Nielsen, R. Molecular Signatures of Natural Selection. *Ann. Rev. Genet.* **39**, 197-218,
1268 doi:10.1146/annurev.genet.39.073003.112420 (2005).
1269
1270 180 Siewert, K. M. & Voight, B. F. Detecting Long-Term Balancing Selection Using Allele
1271 Frequency Correlation. *Mol. Biol. Evol.* **34**, 2996-3005, doi:10.1093/molbev/msx209
1272 (2017).
1273
1274 181 Voight, B. F., Kudravalli, S., Wen, X. & Pritchard, J. K. A Map of Recent Positive
1275 Selection in the Human Genome. *PLoS Biol.* **4**, e72, doi:10.1371/journal.pbio.0040072
1276 (2006).
1277

- 1278 182 Messer, P. W. & Petrov, D. A. Population genomics of rapid adaptation by soft
1279 selective sweeps. *Trends Ecol. Evol.* **28**, 659-669, doi:10.1016/j.tree.2013.08.003
1280 (2013).
1281
- 1282 183 DeGiorgio, M., Lohmueller, K. E. & Nielsen, R. A Model-Based Approach for Identifying
1283 Signatures of Ancient Balancing Selection in Genetic Data. *PLoS Genet.* **10**, e1004561,
1284 doi:10.1371/journal.pgen.1004561 (2014).
1285
- 1286 184 Kim, Y. & Stephan, W. Detecting a local signature of genetic hitchhiking along a
1287 recombining chromosome. *Genetics* **160**, 765-777 (2002).
1288
- 1289 185 Cheng, X. & DeGiorgio, M. Flexible mixture model approaches that accommodate
1290 footprint size variability for robust detection of balancing selection. *bioRxiv*, 645887,
1291 doi:10.1101/645887 (2019).
1292
- 1293 186 Csilléry, K., Blum, M. G. B., Gaggiotti, O. E. & François, O. Approximate Bayesian
1294 Computation (ABC) in practice. *Trends Ecol. Evol.* **25**, 410-418,
1295 doi:10.1016/j.tree.2010.04.001 (2010).
1296
- 1297 187 Schrider, D. R. & Kern, A. D. Supervised Machine Learning for Population Genetics: A
1298 New Paradigm. *Trends Genetics* **34**, 301-312, doi:10.1016/j.tig.2017.12.005 (2018).
1299 **This paper offers an accessible description of both present applications and possible**
1300 **future developments of supervised machine learning for understanding signatures of**
1301 **selection in genomic scale data.**
1302
- 1303 188 Raynal, L. *et al.* ABC random forests for Bayesian parameter inference. *Bioinformatics*
1304 **35**, 1720-1728, doi:10.1093/bioinformatics/bty867 (2018).
1305
- 1306 189 Rasmussen, M. D., Hubisz, M. J., Gronau, I. & Siepel, A. Genome-Wide Inference of
1307 Ancestral Recombination Graphs. *PLoS Genet.* **10**, e1004342,
1308 doi:10.1371/journal.pgen.1004342 (2014).
1309
- 1310 190 Kelleher, J., Etheridge, A. M. & McVean, G. Efficient Coalescent Simulation and
1311 Genealogical Analysis for Large Sample Sizes. *PLOS Computational Biology* **12**,
1312 e1004842, doi:10.1371/journal.pcbi.1004842 (2016).
1313
- 1314 191 Haller, B. C., Galloway, J., Kelleher, J., Messer, P. W. & Ralph, P. L. Tree-sequence
1315 recording in SLiM opens new horizons for forward-time simulation of whole genomes.
1316 *Mol Ecol Resour* **19**, 552-566, doi:10.1111/1755-0998.12968 (2019).
1317 **This paper describes the implementation of tree-sequence recording into the already**
1318 **multifaceted and powerful SLiM simulation framework, and provides one of the most**
1319 **important schemes needed to model neutral and non-neutral dynamics on genome-**
1320 **scale data.**
1321
- 1322 192 Hejase, H. A., Dukler, N. & Siepel, A. From Summary Statistics to Gene Trees: Methods
1323 for Inferring Positive Selection. *Trends Genetics*, doi:10.1016/j.tig.2019.12.008 (2020).
1324 **This is an exceptionally comprehensive review of both historical and present**
1325 **approaches for detecting forms of positive selection. While the focus is on positive**
1326 **selection, many of the focal methodologies would, with some modification, be**
1327 **applicable for detecting the many signatures of host-parasite coevolution.**

1328 **Acknowledgements**
 1329 We thank the Ebert lab for fruitful discussion and Suzanne Zweizig for comments on the language of
 1330 the manuscript. This work is supported by a grant from the Swiss National Science Foundation.

1331 **Author contributions**
 1332 The authors contributed equally to all aspects of the article.

1333 **Competing interests**
 1334 The authors declare no competing interests.

1335
 1336
 1337

1338 **Table 1. Comparison of features of specific coevolution by selective sweep and**
 1339 **balancing selection.**

1340

Feature	Selective sweep coevolution	Balancing selection coevolution
Form of selection	Positive selection drives sweeps; selection is directional.	Negative frequency-dependent selection gives common alleles a disadvantage; selection results in a balance of the frequencies of genetic variants.
Functional polymorphisms	Visible only during selective sweeps.	Maintained constantly and potentially for very long time periods.
Underlying genetic system	Beneficial mutation in host and parasite at any locus in the nuclear or cytoplasmic genome may sweep.	Frequencies of alternative alleles at a few selected loci are balanced.
Role of mutations	Mutations define the onset of new selective sweeps (hard sweeps).	Mutations are not necessary but do create rare variants, which may be selected and contribute to balancing selection or even replace a previous variant.
Temporal continuity	Process can be highly stochastic and does not need to be continuous; long periods without sweeps are possible.	Process must operate continuously because genetic variants may otherwise be lost. In a spatial setting, previously lost alleles may be reintroduced from other populations.
Time scale of phenotypic change	Relatively slow because new mutations take a long time to reach a high enough frequency to be recognized. Sweeps starting from standing genetic variation progress more quickly.	Fast because genetic variants are always at intermediate frequencies where selection results in fast changes.
Population divergence	Sweeps drive population and species divergence.	Population divergence is prevented in the long term, though it may occur in the short term.
Evolutionary outcome	Creates macroevolutionary patterns (lineage divergence).	Explains high levels of genetic diversity within populations and species.
Introgression among species	May introduce beneficial new alleles that can sweep.	May introduce new functional variants that can contribute to balancing selection, but may create a fake picture of trans-species polymorphism.

1341

1342
 1343
 1344
 1345
 1346
 1347
 1348
 1349
 1350
 1351
 1352
 1353

Table 2. Genomic summaries used to determine the evolutionary process underlying a genomic signature.

Signature type refers to the two models of specific coevolution that may be supported with this summary. Time-scale refers to the approximate period during which the summary is able to detect a signal for the genomic signature relative to the genomic background (compare Fig. 2). CLR, Composite Likelihood Ratio Test; HKA, Hudson, Kreitman and Aguadé test; iHS, integrated haplotype score; LD, linkage disequilibrium; MK, McDonald-Kreitman; NCD, Non-central deviation; SFS, site frequency spectrum; TSP, trans-species polymorphism.

Name of summary	Type of summary	Description	Signature type (relative to genomic background)	Time-scale	Local genomic scale of analysis	Other comments
Allele vs. species-tree discordance ¹⁷⁵	Summary statistic	Compares phylogenetic relationships of genes to those of species.	Discordant trees indicate balancing selection and/or TSP	Deep	Distinct genes and/or windows across the genome	–
Elevated ratio of adaptive to non-adaptive divergence ^{176,177}	Summary statistic and likelihood method	Measures the ratio of adaptive to synonymous divergence	High values indicate positive selection	Deep-to-intermediate	Gene classes can be compared	Explicit tests arise from the MK test or its derivatives
HKA_{trans} ¹⁷⁸	Summary statistic	Adaptation of the standard HKA test to better accommodate genomic data from multiple species	Positive values of chi-squared test statistic are suggestive of balancing selection	Deep-to-intermediate	Windows across genome	–
NCD_{trans} ^{28,178}	Summary statistic	Extension of the NCD summary statistic to accommodate multi-species data	Increasing values are suggestive of balancing selection	Deep-to-intermediate	Windows across genome	See also ¹⁷⁸ . Describes the NCD1 summary which includes an outgroup and fixed differences as part of the test
Tajima's D and HKA ¹⁷⁹	Summary statistic	Excess of low frequency class polymorphisms	Decreased values indicate positive selection/ Selective sweep	Intermediate-to-recent	Windows across genome	Summary can be confounded with demography due to population expansion causing similar measures of the summary statistic
		Increased values indicate balancing selection	Increased values indicate balancing selection	Deep-to-intermediate	Windows across genome	Summary can be confounded with demography due to population bottlenecks

						causing similar measures of the summary statistic
Perturbations from equilibrium SFS ¹⁷⁹	Summary statistic and likelihood method	Intermediate-skewed allele frequency classes	Excess derived allele frequency classes as compared to neutral expectation is indicative of balancing selection	Deep-to-intermediate	Windows across genome	Summary can be confounded with demography due to population bottlenecks causing similar measures of the summary statistic
		Excess minor and/or derived allele frequency classes	Excess minor or derived allele frequency classes as compared to neutral expectation is indicative of positive selection and selective sweeps	Intermediate-to-recent	Windows across genome	Summary can be confounded with demography due to population expansion causing similar measures of the summary statistic
β ¹⁸⁰	Summary statistic	Measure of allele frequency correlation and overall mutation rate.	Values greater than 0 are indicative of balancing selection	Intermediate	Windows across genome	
iHS ¹⁸¹	Summary statistic	Lengths of haplotype homozygosity or the decelerated decay of LD	Increased lengths of homozygosity and the slower decay of LD compared to neutral expectations is indicative of a selective sweep	Recent	Windows across genome	
LD ¹⁷⁹	Summary statistic	Distinct increase in LD amongst a subset of adjacent loci	Elevated values above genome wide background indicate a selective sweep	Recent	Windows across genome	
H₁, H₁₂, H₂ ^{42,182}	Summary statistic	Frequencies of ranked haplotypes, that is haplotype spectra	Increased frequencies of distinct, ranked haplotypes (and their relative frequencies)	Deep-to-intermediate	Windows (SNPs) across genome	

			is indicative of a sweep (hard versus soft).			
T_{trans} ¹⁷⁸	Likelihood method	Adaptation of the T statistic of ¹⁸³ to specifically detect TSP	Fit of the observed data to a model of long-term balancing selection or TSP is greater than that of a model of neutral evolution (CLR _{TSP} > CLR _{Neutral})	Deep-to-intermediate	Windows across genome	
T statistic ¹⁸³	Likelihood method	Aggregate of summary statistics of genomic diversity parameterize a maximum likelihood-based comparison of the fit of a neutral vs. balanced polymorphism model	Fit of the observed data to a model of balancing selection is greater than that of a model of neutral evolution (CLR _{Balancing selection} > CLR _{Neutral})	Intermediate-to-recent	Windows across genome	
CLR ¹⁸⁴	Likelihood method	Aggregate of summary statistics of genomic diversity parameterize a maximum likelihood-based comparison of the fit of a neutral vs. selective sweep model	Fit of the observed data to a model of balancing selection is greater than that of a model of neutral evolution (CLR _{Sweep} > CLR _{Neutral})	Recent	Windows across genome	Refers specifically to the model described in ref. ¹⁸⁴ and its derivatives; while CLR is used sometimes to refer to this specific summary, composite likelihood ratio tests are a general statistical procedure for model comparison.

1354
1355
1356

1357 Figure legends

1358

1359

1360 Figure 1 | **Schematic representation of selective sweep coevolution in the gene pools**
1361 **of a host and parasite population.** For each antagonist, timelines of alleles at different
1362 loci are shown in different colours. Over the shown time period, each of the alleles is
1363 replaced at least once by a mutant, indicating selective sweeps. Genetic variation at
1364 loci, that is the presence of multiple alleles in the gene pool at a given moment in time,
1365 is only visible during the sweep. Host and parasite sweeps are not linked. Loci can be
1366 present anywhere in the genome. Other loci may undergo sweeps at the same time for
1367 reasons unrelated to the coevolutionary interactions.

1368

1369

1370 Figure 2 | **The temporal dynamics of genomic signatures.**

1371 Approximate time scales during which genomic signatures for balancing selection
1372 (blue), positive selection and/or selective sweep (red), and interspecies linkage
1373 disequilibrium (purple) are detectable with common (though not exhaustive)
1374 population genetic tests (Table 2). While signatures of balancing selection (especially
1375 those that are associated with trans-species polymorphism (TSP) or approximations
1376 thereof; see ¹⁸⁵) and positive selection may extend quite deep into evolutionary time,
1377 patterns of very rapid evolutionary change associated with positive selection or
1378 selective sweeps will show a much more limited time depth. Concomitantly, the
1379 number of generations a given selective regime has to be in place before inference of
1380 its signature is possible may be different between balancing selection and positive
1381 selection and, on the shortest times scales and with a sample from a single population,
1382 the signatures of balancing selection and positive selection may be indistinguishable.
1383 As a selective sweep grows older, its signature will disappear and become more similar
1384 to the signature of positive selection, which is visible only as elevated ratios of
1385 adaptive to synonymous divergence. Patterns of interspecies linkage disequilibrium
1386 provide no historical record of the coevolutionary process and so additional methods
1387 (such as those described in this figure) are required to determine what (if any)
1388 selective regime has been at play at the statistically associated loci. It should be noted,
1389 the suggested time frame of inference for individual tests may differ substantially for
1390 individual datasets as a result of experimental factors, such as sample size, and as the
1391 result of species-specific factors, such as demography. Allele frequency classes may be
1392 a component of summary statistic (such as Tajima's D) or may be the specific metric of
1393 interest (for example, site-frequency spectra, or SFS). T_{TRANS} = likelihood ratio test
1394 statistics for detecting TSP; HKA_{TRANS} = modified version of the HKA test to
1395 accommodate genomic data from multiple species in order to detect TSP; NCD = non-
1396 central deviation summary to determine deviation from neutral allele frequency
1397 expectations; H_1, H_2 , and H_{12} = Frequencies of the most frequent haplotype, the
1398 frequency of the second most frequent haplotype, and the frequency of the first and
1399 second most frequent haplotype, respectively; NCD_{TRANS} = version of the NCD summary
1400 statistic extended to multispecies datasets in order to detect TSP ^{28,178}; HKA = Hudson,

1401 Kreitman and Aguadé test; SFS = site frequency spectrum; iHS = integrated haplotype
1402 score ¹⁸¹; LD = linkage disequilibrium ¹⁷⁹; CLR = composite likelihood ratio test ¹⁸⁴; β =
1403 measure of allele frequency correlation ¹⁸⁰ (see Table 2).

1404
1405

1406 **Figure 3 | Host resistotype and parasite infectotype interaction matrix and balancing**
1407 **selection.** A simple matching allele model of two host resistotypes and two parasite
1408 infectotypes is sufficient to create long term balancing selection within populations. **a**
1409 | Haploid hosts with the H-allele have resistotype C (C for compatible interactions, that
1410 is, the host is susceptible) for parasites with the P-allele and I for incompatible (the
1411 host is resistant) for parasites with the p-allele; hosts with the h-allele have resistotype
1412 I for parasites with the P-allele and C for parasites with the p-allele. The haploid
1413 parasite with the P-allele has the infectotype C for hosts with the H allele and I for
1414 hosts with the h-allele, whereas parasites with the p-allele have infectotype I for hosts
1415 with the H-allele and C for hosts with the h-allele. A mutant of an allele may produce a
1416 new allele (H', h', P' and p'), but if the function of the allele is not affected in regards
1417 to its interaction with the parasite, the resistotypes and infectotypes will stay the
1418 same. **b** | A schematic representation of balancing selection is shown in a host
1419 population that diverges into two populations. Allele colours correspond to functional
1420 types of the host. Initially, the H-allele (blue) gives rise to the h-allele (red), which has a
1421 different resistotype. Afterwards, the two allele types are maintained by balancing
1422 selection. Their infection profile might resemble the matching allele model shown in
1423 part a. Further mutations in the H- and h-alleles change the genotype of these alleles,
1424 but not their resistotype. The H-allele persists until it is replaced by the H' allele (light
1425 blue) in population 1. The h-allele gives rise to the h' allele (pink) in population 2. H
1426 and H', as well as h and h', change in relative frequency independent of selection by
1427 the parasite. Gene flow between populations can introduce new alleles (h' into
1428 population 1) and reintroduce extinct alleles (h' into population 2), which can replace
1429 the resident allele.

1430
1431

1432 **Figure 4 | An idealized scenario for host–parasite coevolution by long-term balancing**
1433 **selection leading to trans-species polymorphisms.**

1434 **a** | When balancing selection maintains allelic variants over long time periods, trans-
1435 species polymorphism (TSP) may be visible, with the polymorphisms existing prior to
1436 the split into two species. In contrast to the scenario in FIG. 3b, here, speciation
1437 completely blocks gene flow and hybridization. If TSP results from strict long-term
1438 coevolution, similar evolutionary histories are expected for the functionally linked
1439 variants in hosts and parasites (FIG. 3a). **b** | For the scenario outlined in part a, the
1440 genes undergoing long-term balancing selection are expected to cluster by function in

1441 a gene tree (red and blue colour), not by species. Host and parasite genealogies
1442 should, however, be congruent (that is, have the same topology), indicating that
1443 speciation events occurred in parallel.

1444
1445

1446 **Figure 5 | Cogenomic approaches to find genes involved in host parasite interactions.**

1447 A cogenomic analysis tests for associations between polymorphisms in host and
1448 parasite genomes. **a** | The genome-to-genome method analyses samples of hosts
1449 naturally infected with a parasite to find host variants in strong interspecies linkage
1450 disequilibrium (iLD) with variants in the parasites. Phenotypic data are not necessary.
1451 The figure shows two pairs of sites in the host and parasite genomes that significantly
1452 associate with each other, that is, they show strong iLD. **b** | The analysis with a Two-
1453 Organism Mixed Model (ATOMM) includes phenotypic data from a host-parasite
1454 infection matrix, which are analysed together with polymorphism data for both the
1455 host and parasite isolates ¹⁶⁴. The method allows interacting loci in hosts and parasites
1456 to be detected. Host and parasite genomes are represented as a series of squares,
1457 where each square indicates the position of a polymorphism in the sampled genomes.
1458 Shades of blue and red indicate different host and parasite genomes, respectively.
1459 Unequal numbers of host and parasite isolates can be used.

1460
1461
1462

1463 **Box 1. Emerging cogenomics methodologies**

1464

1465 Likelihood-free inference

1466 The biological complexity inherent to the coevolutionary process and the distinct
1467 limitations of population genetic summary statistics, including the lack of a suitable
1468 and/or tractable **likelihood function** [G], for dealing with certain aspects of this
1469 complexity have driven researchers to seek new solutions¹⁴⁹. A very successful
1470 approach for by-passing some limitations of individual summary statistics and
1471 likelihood-based inference has been Approximate Bayesian Computation (ABC). In
1472 ABC, the requirement of a likelihood function is bypassed by the use of simulated
1473 datasets. Summarization of these simulations via summary statistics followed by a
1474 statistical evaluation, or rejection procedure, identifies which simulations most closely
1475 approximate the set of summary statistic values obtained from an observed dataset,
1476 thereby allowing inference of the posterior probabilities of model parameters of
1477 interest¹⁸⁶. Another exciting approach that allows the aggregate use of simulation and
1478 summary statistics to evaluate complex biological dynamics is **supervised machine**
1479 **learning** [G]. It works on the principle of using training sets to generate a predictive
1480 model, for example a simulated dataset of the relevant evolutionary scenarios to a
1481 given problem of interest. This predictive model is then used to determine how a set
1482 of input variables predicts a given response¹⁸⁷. However, these different approaches
1483 should not be considered mutually exclusive¹⁸⁸.

1484

1485 Ancestral recombination graphs

1486 To identify a signature of selective sweep or balancing selection using the methods
1487 described above, a clear understanding of how a neutral coalescent differs from a non-
1488 neutral one is crucial. Given that so much useful information can be gleaned from the
1489 flanking genomic regions as well as the locus experiencing selection, traditional
1490 summaries of the coalescent that rely on treating loci or genomic windows as (semi-)
1491 independent will throw away important, if not crucial, information needed to infer the
1492 dynamics of the coevolutionary process. The **ancestral recombination graph** [G] (ARG)
1493 provides a representation of the relationships among genomic segments, mediated
1494 through recombination, as a network¹⁸⁹. Historically, it has been computationally
1495 prohibitive to reconstruct these ARGs in even small genomic data sets. However,
1496 recent advances in the data structures required to encode the information of an ARG
1497¹⁹⁰, and in the simulation of complicated, non-neutral perturbations of the coalescent
1498 in the context of whole genomes¹⁹¹, have begun to allow the inference of both
1499 selective sweeps and balancing selection from reconstructed ARGs^{189,192}. It is perhaps
1500 not inconceivable that future advances may well allow for the direct reconstruction of
1501 the pairwise relationships of genomic segments in host and parasite, which is
1502 mediated through the reproductive processes of host and parasite, and the infection
1503 process that connects them.

1504

1505 **Glossary**

1506

1507

1508 PARASITES

1509 Organisms, including pathogens, that take advantage of other organisms (hosts),
1510 thereby instigating a process of selection by the host to defend against the parasite.

1511

1512 GENOMIC SIGNATURES

1513 Characteristic patterns of genetic variation, observed at a genomic region in a sample
1514 of genomes.

1515

1516 SELECTIVE SWEEP

1517 The spread of a beneficial mutant and the hitchhiking of genetic variants close to it in
1518 the genome. Beneficial mutants may have arisen de novo or were segregating in the
1519 population before the sweep and become beneficial after a change in conditions.

1520

1521 LINKED SELECTION

1522 The evolution of (nearly) neutral SNPs influenced by selection on loci physically linked
1523 to them.

1524

1525 GENETIC HITCHHIKING

1526 The process by which a genetic variant changes in frequency because it is physically
1527 linked to another variant that is changing in frequency due to selection.

1528

1529 COGENOMIC

1530 The simultaneous analysis of the function, structure and evolution of pairs of
1531 associated genomes in closely-interacting organisms, such as host and parasites.

1532

1533 BALANCING SELECTION

1534 Balancing selection occurs because the alleles involved have, on average, a selective
1535 advantage that correlates negatively with their frequency within a population or
1536 species. This term does not include loci that are at a balance between gaining and
1537 losing variants, such as mutation-drift and mutation-selection balance.

1538

1539 OVERDOMINANCE

1540 Describes the scenario in which heterozygotes have a more extreme phenotypic trait
1541 value than all homozygotes. Overdominance for fitness results in balancing selection
1542 for the alleles causing the advantage for the heterozygote genotypes.

1543

1544 LOCAL ADAPTATION

1545 If the effect of an allele is habitat specific, such that it is beneficial in one habitat and
1546 detrimental in another and vice versa for the alternative allele, local adaptation may
1547 evolve with directional selection within each population. Local adaptation is a powerful
1548 mechanism to maintain genetic diversity within species.

1549

1550 DIRECT NEGATIVE FREQUENCY-DEPENDENT SELECTION
1551 Occurs when the selective benefit of an allele depends directly and negatively on its
1552 frequency, for example at sex-determining loci and plant self-incompatibility loci.
1553
1554 INDIRECT NEGATIVE FREQUENCY-DEPENDENT SELECTION
1555 Occurs when the selective benefit of an allele depends on the frequency of an allele in
1556 a coevolving species.
1557
1558 SELECTIVE INTERFERENCE
1559 In clonal, but not in sexual, populations beneficial mutations interfere with each other,
1560 such that at a given moment the fittest mutation will outcompete weaker beneficial
1561 mutations. Interference can also affect the spread of mutants in genomic regions with
1562 low recombination rates.
1563
1564 LINKAGE DISEQUILIBRIUM (LD)
1565 A statistical measure of the distribution of combinations of alleles at different loci,
1566 which is zero if this distribution follows the expectation based only on allele
1567 frequencies. Non-zero values of LD can arise due to hitchhiking, selection on allele
1568 combinations and stochastic processes and may occur among loci without physical
1569 linkage.
1570
1571 SELECTION COEFFICIENT
1572 A measure of fitness of genotypes or alleles relative to a reference, such as the
1573 ancestral form.
1574
1575 TAJIMA'S D
1576 A population genetic summary statistic describing the frequency distribution of
1577 polymorphisms in a population, with D being zero under neutral evolution and positive
1578 under balancing selection.
1579
1580 GENETIC DRIFT
1581 A neutral evolutionary process that influences allele frequencies based on the random
1582 sampling of genetic variants during reproduction.
1583
1584 PANMICTIC
1585 Random mating within a population.
1586
1587 FIXATION INDEX (F_{ST})
1588 A measure of genetic differentiation of spatially structured populations, usually
1589 estimated from single-nucleotide polymorphism (SNP) or microsatellite data.
1590
1591 DIRECTIONAL SELECTION
1592 A mode of natural selection by which a genetic variant is predicted to spread to
1593 fixation (also known as positive selection).
1594
1595

1596 SUPERGENE
1597 A group of tightly linked genes on a chromosome that are inherited together as a
1598 haplotype and often have reduced recombination.
1599

1600 FUNCTIONAL GUILDS
1601 Groups of organisms with similar life-style characteristics that perform the same
1602 ecological function, such as gut-parasites, pollinators and filter-feeders.
1603

1604 R-GENE
1605 Resistance genes of plants that convey resistance against diseases by producing R
1606 proteins.
1607

1608 APPROXIMATE BAYESIAN COMPUTATION (ABC)
1609 Bayesian statistical approach wherein parameter inference and model selection is
1610 conducted in the absence of likelihood functions. Instead, ABC relies on summary
1611 statistics and simulations to infer posterior distributions of parameters and/or models
1612 of interest.
1613

1614 LIKELIHOOD FUNCTION
1615 Analytical formulation of a set of parameters, which can be used to assess the fit of a
1616 given observed dataset to a predetermined model.
1617

1618 SUPERVISED MACHINE LEARNING
1619 Machine learning is a statistical methodology that uses artificial intelligence to
1620 automate inferential processes with minimal explicit instruction. Supervised machine
1621 learning is a type of machine learning that uses (labelled) training sets to generate a
1622 target function when the correspondence between the function of interest and
1623 response variable are known. This target function can then be applied to unclassified
1624 (unlabelled) data to make statistical inferences.
1625

1626 ANCESTRAL RECOMBINATION GRAPH (ARG)
1627 A genealogical or phylogenetic representation of the network of coalescence and
1628 recombination events in a collection of orthologous DNA sequences.
1629

1630 **ToC summary**
1631 Host-parasite co-evolution is expected to leave signatures of selection in the genome
1632 of both antagonists. Ebert and Fields discuss what is known about these signatures,
1633 how they relate to co-evolutionary processes, and how they can help identify the
1634 genes underlying the coevolving phenotypes.
1635
1636

Figure 1.

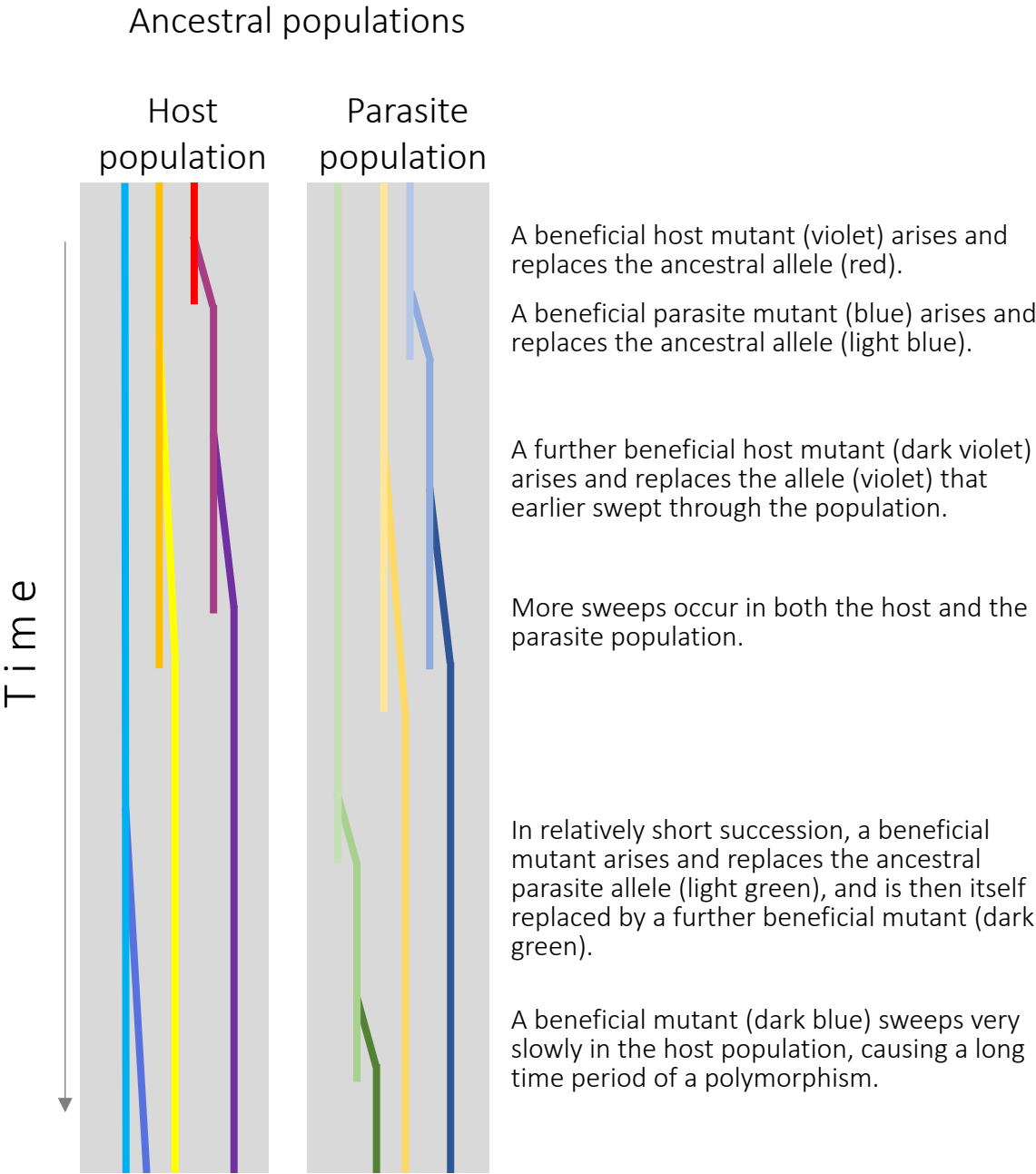


Figure 2.

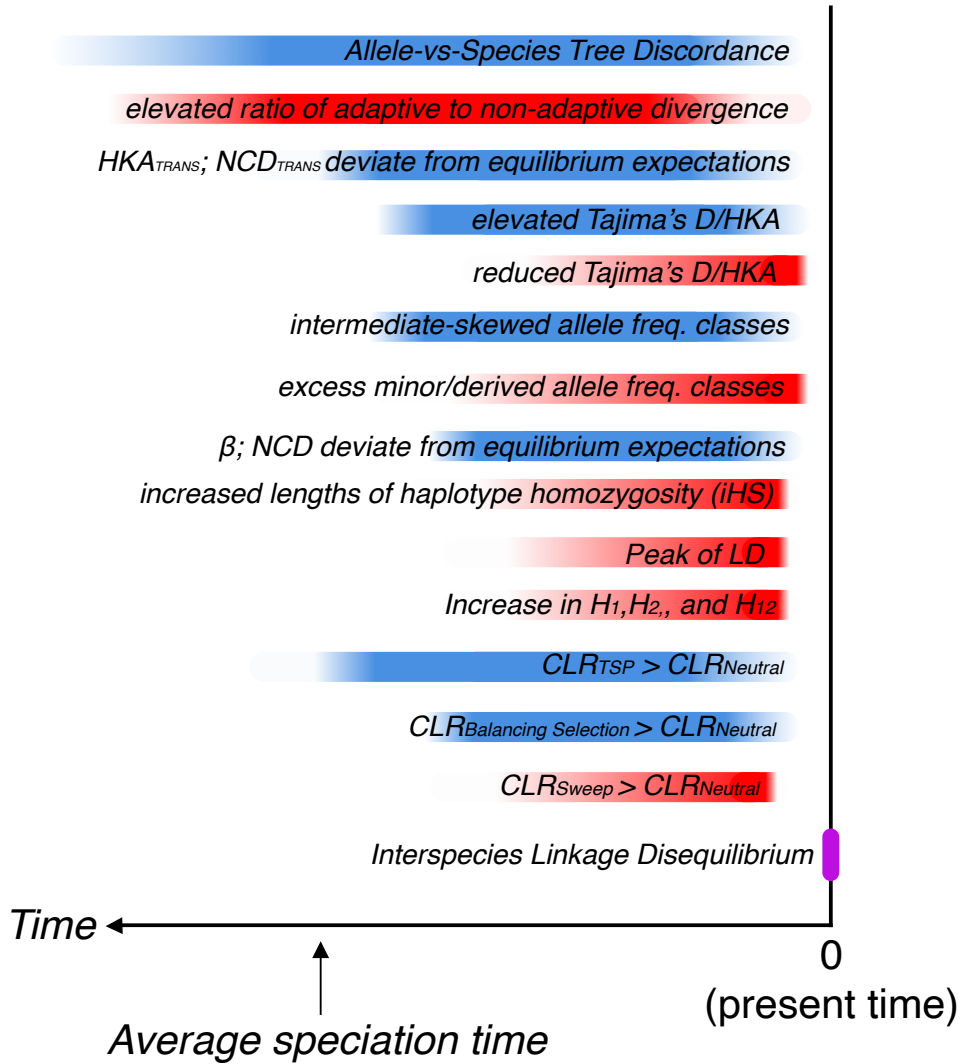


Fig. 3.

a

		Parasite alleles	
		B, B'	b, b'
Host alleles	A, A'	compatible, C	incompatible, I
	a, a'	incompatible, I	compatible, C

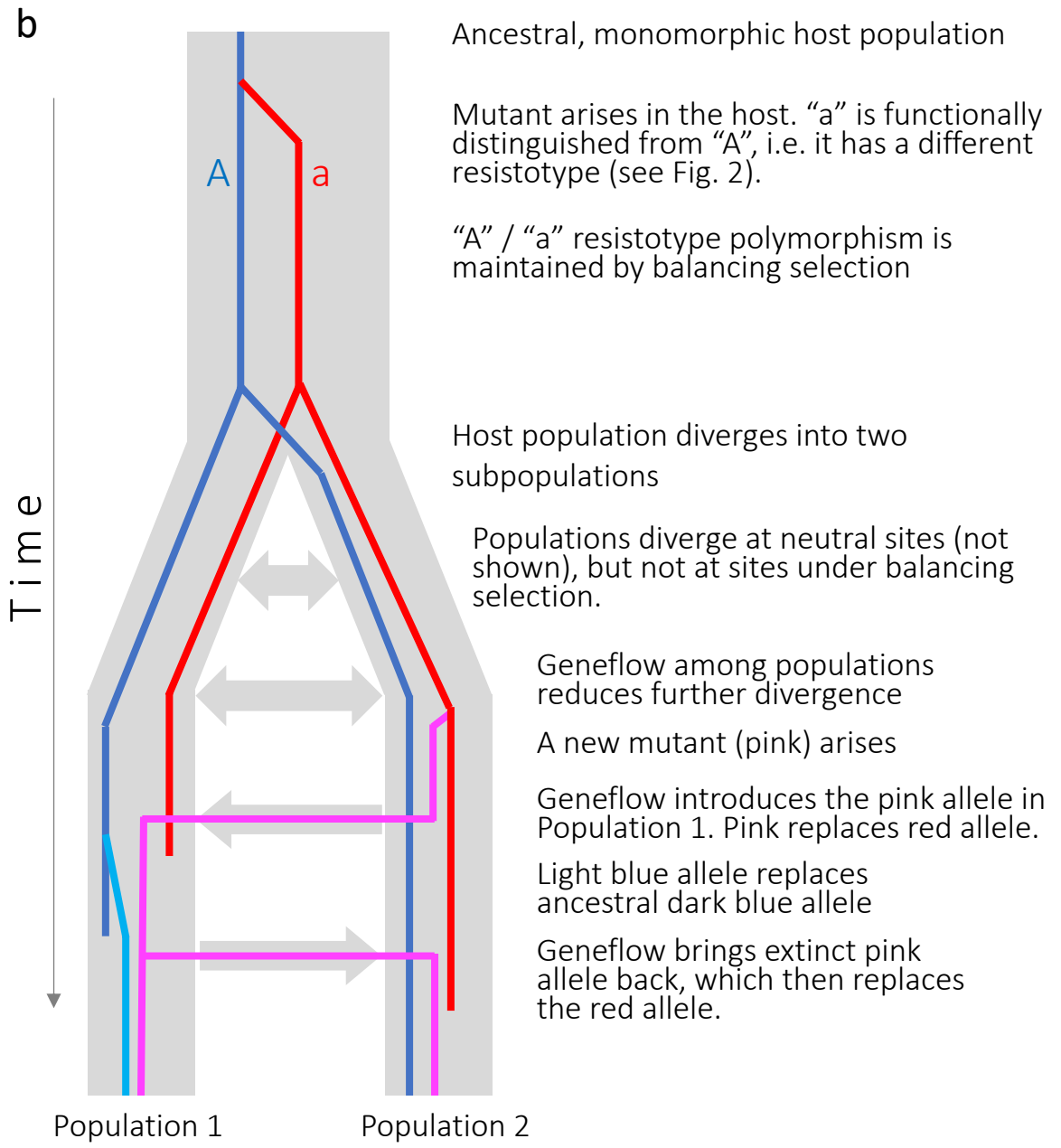


Fig. 4.

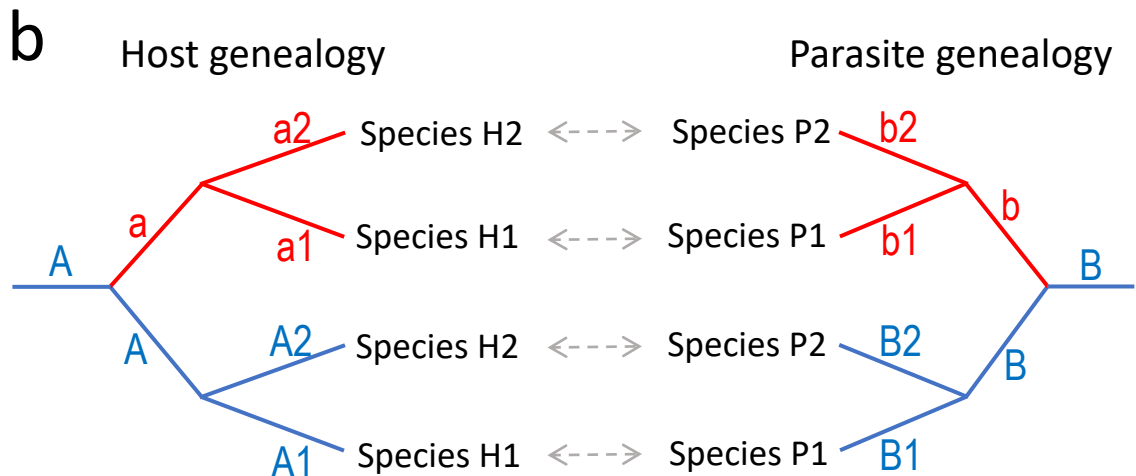
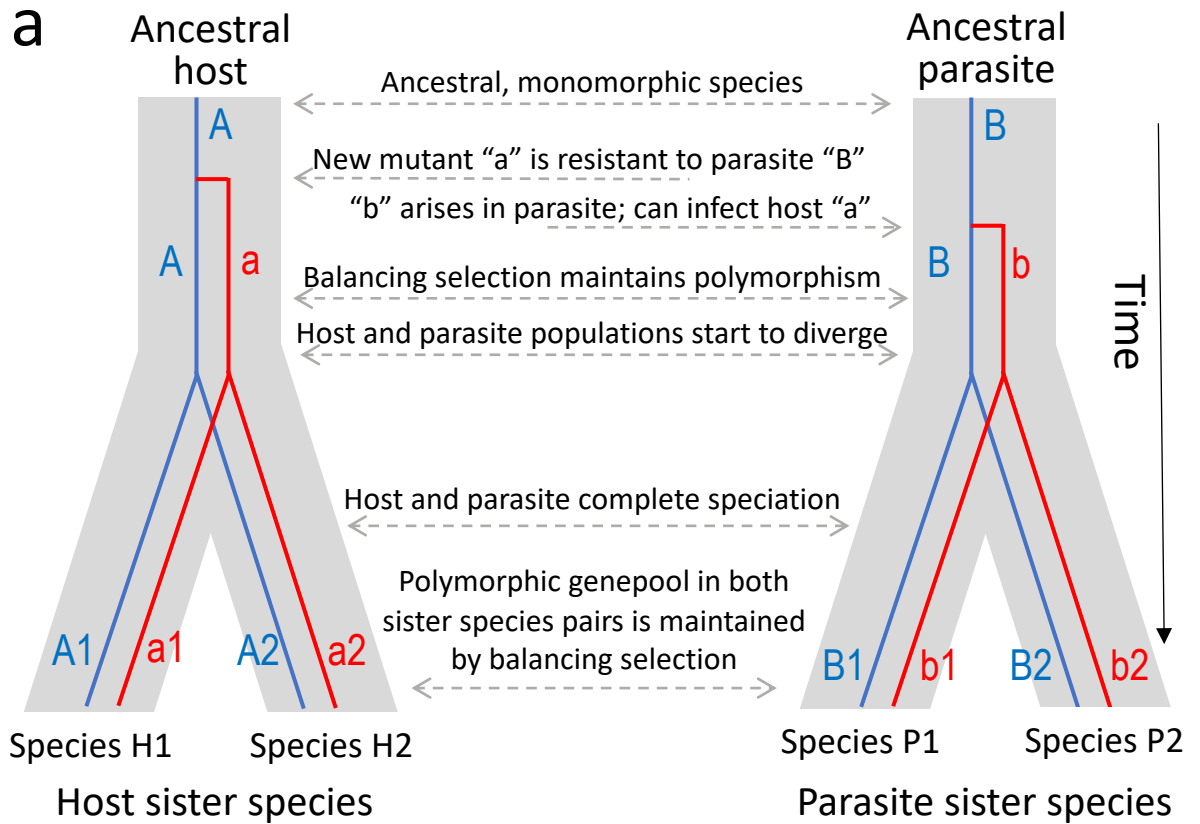


Fig. 5

