

Beyond genes-for-behaviour: the potential for genomics to resolve questions in avian brood parasitism

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Abstract

Brood parasite-host interactions are among the most easily observable and amenable natural laboratories of antagonistic coevolution, and as such have intrigued evolutionary biologists for decades. It is therefore surprising they have not been at the forefront of genomic studies on evolutionary adaptation. Here we review state-of-the-art molecular methods in studying avian brood parasitism, a model system in behavioural ecology. We highlight outstanding questions to bring examples of how genomic tools are not merely about ‘finding a gene for behaviour’, but can be used to study the causes and mechanisms of (co)evolutionary adaptation. In doing so, we promote behavioural and molecular ecologists to integrate Tinbergen’s questions into a collaborative, coherent science aiming to solve the mysteries of nature and apply current methodology into other model systems in behavioural ecology.

Behavioural ecology and genomics

The need to close the gap in knowledge between phenotype and genotype has long been discussed [1]. However, reading genome sequences has been insufficient to explain phenotypic variation [2–4] and has instead revealed that closing the gap requires understanding links between the genotype, phenotype, environment and species interactions across biological levels (e.g. [5,6]). Consequently, recognising when and how genomic data may be useful to study the ecology and evolution of behaviour is yet to progress beyond human genome pioneer

32 Eric Lander's famous quote: "Genome: Bought the book; hard to read". Bridging behavioural
33 and molecular ecology could however bring a major advance. Behavioural ecology provides a
34 rich understanding of why behaviour and associated traits (e.g. morphology and physiology)
35 evolve at the phenotypic level and a tool-kit of quantitative and experimental methods [7] to
36 address how the actions of individuals influence population-level processes in the wild [8].
37 However, behavioural ecology continues to work mostly at the level of the phenotype ([figure](#)
38 [1a](#)), and is sometimes viewed as the 'soft underbelly' of evolutionary biology [8]. In contrast,
39 the declining costs of next-generation sequencing ([figure 1a](#)) have helped molecular ecologists
40 use genomic data to identify genes and traits relevant for fitness and adaptation [9,10] including
41 in non-model organisms [11]. Behaviour, however, is rarely considered in molecular ecology,
42 despite it influencing heritability (i.e. non-genetic inheritance and indirect genetic effects [12])
43 and being one of the major drivers determining how genes interact in time and space [13] and
44 experience selection.

45

46 While we are not the first to call for the use of genomic tools to study behaviour (e.g.[14–18],
47 they are still rarely adopted in behavioural ecology ([figure 1a](#)) despite earlier genetic
48 approaches leading to major advances (e.g. microsatellites to determine parentage [19]). This
49 might be because 'finding the gene/s for a behaviour' is often the focus (e.g.[14,15]) and this
50 is rarely feasible or cost effective in the wild [18] (See box 1). However, genomic tools can be
51 used to do more than uncover the mechanistic basis of traits; they can help us find answers to
52 questions about trait function, evolutionary history and development (i.e. Tinbergen's Four
53 questions [20]). For example, (i) whole-genome sequencing data can help improve the
54 robustness of phylogenetic trees and comparative analyses, by resolving discrepancies between
55 different loci [21]; (ii) by comparing neutral and selected markers across genomes, we can now
56 infer population demographics and local adaptation in addition to population structure and gene
57 flow [22]; and (iii) genome sequencing provides increased resolution to determine parentage
58 [23] and facilitates inferring offspring (or hybrid) fitness (through pedigree reconstruction,
59 e.g.[24]). Perhaps most importantly, (iv) genomics allows us to go further than just improving
60 on previous genetic methods: we can acquire information about species' evolutionary history
61 and adaptive potential from the genome [25,26] to predict impacts of future change [27]. This
62 is now a critical question for everyone working in ecology and evolution.

63

64 One of the largest stumbling blocks to integration is a cultural divide: while both behavioural
65 and molecular ecologists study adaptation and address both ultimate "why" and proximate

66 “how” questions, the two fields are separated by different approaches, jargon (see table 1 for
67 examples) and working styles. Behavioural ecology was founded on the ‘phenotypic gambit’,
68 where knowing heritable genetic mechanisms is considered unnecessary as long as the trait
69 correlates with fitness proxies [28]. Meanwhile, to resolve technological and methodological
70 issues inherent to studying the heritability of polygenic traits (like behaviour), molecular
71 ecologists have largely focused on those thought easier to measure (e.g. morphology [29]) and
72 assumed to be determined by few, large-effect loci [26]. However, it is becoming evident that
73 this approach has biased conclusions, and a single gene or supergene is unlikely to be a
74 common explanation for variation across phenotypic traits [18], including in model systems
75 (e.g. chicken plumage and egg coloration [30]). Furthermore, while heritability estimates for
76 behavioural traits (when investigated) are often within range of estimates for physiological and
77 life-history traits [29], recent genomic studies suggest that mechanisms of heritability are more
78 complex than previously thought (see e.g.[3] for epigenomics and regulatory mechanisms) and
79 the genome is now considered an environmentally responsive entity [6] somewhat analogous
80 to behaviour [14].

81

82 Massive genomic datasets for species where we already have (or have the potential to collect)
83 rich behavioural data are becoming available (e.g. Avianbase [31]), so how can we overcome
84 this cultural divide? Here we bring together behavioural and molecular ecologists (box 2) to
85 work through a case study example and demonstrate how we can combine information from
86 both the genome and behaviour to study adaptation. We use one of the most classic textbook
87 examples in behavioural ecology (avian brood parasitism [32]) and assess what is (and what is
88 not) possible with the rapidly expanding, and sometimes overwhelming [33–35], range of
89 genomic tools and analytical methods available [1,36] to address broad questions of interest.

90 Avian brood parasitism as a case study

91

92 Obligatory brood parasites (about 1% of birds, some insects and the cuckoo catfish [32]) trick
93 hosts into rearing foreign offspring as their own. This selects for the evolution of host defences,
94 with counter-adaptations in parasites [37] ([figure 2](#)) and has provided behavioural ecologists
95 with a particularly tractable system for studying coevolution in the wild [38]. Experimental
96 methods to test hypotheses about behavioural defences and cuckoo trickery [39] are well
97 defined, and techniques to measure phenotypes are taking advantage of technological

98 developments (e.g. egg pattern and colour [40] to identify individual cuckoos [41]. Compared
99 to the rich body of knowledge for behavioural adaptations, almost nothing is known about the
100 molecular mechanisms underpinning behavioural defences and offences and the heritability of
101 traits remains mostly assumed. The field of genetics has, however, already provided some
102 insight into the evolutionary history of brood parasitism [42] and host-race specialization [43–
103 45]. For example, microsatellites have been used to estimate levels of gene flow between
104 parasitised and unparasitised magpie populations [46,47], leading to a candidate marker for its
105 egg rejection behaviour [46]. Nevertheless, the same marker was not associated with egg
106 rejection in another host species (the great reed warbler, [48]) and this line of inquiry has not
107 continued.

108
109 Despite repeated predictions that genomics would re-revolutionize the study of brood
110 parasitism (e.g. [17,49,50]), there has been relatively little work making use of data or insights
111 from high-throughput sequencing ([figure 1](#)). The few examples are recent and limited to a small
112 number of species (discussed in more detail below): improved phylogenies [51,52], population
113 structure of brood-parasitic sister-species [53], adult parasite diet [53,54] and nestlings'
114 microbiota [55], the heritability of parasite and host egg coloration [55,56], and the loss of
115 parental care in brood parasites [57]. This is surprising, since biologists working on birds were
116 instrumental in developing the field of quantitative genetics with an animal model approach
117 [58], and after entering the genomic era [59] they quickly generated a detailed understanding
118 of avian genome structure [60] and conducted some of the first studies of genomic adaptation
119 in wild populations (e.g. [27]; [26]). Furthermore, high-quality reference genomes are now
120 available for brood parasite hosts with rich behavioural data (e.g. superb fairy wren: [61], great
121 reed warbler: [62,63], reed warbler: [64]). Therefore, here we discuss how applying genomic
122 methods to avian brood parasitic systems could be used to better predict (i) where and when
123 brood parasitism should evolve, (ii) when and how hosts defend, or (iii) how coevolutionary
124 trajectories depend on ecological change, three major questions in a field aiming for a deeper
125 understanding of coevolutionary dynamics [32]. Even though we focus on long-standing
126 questions in avian brood parasitism, these three key concepts (evolutionary origins, plasticity,
127 geographic selection mosaics) are of broad relevance across ecology and evolution.

129 1) Where and when should brood parasitism evolve?

130 Darwin [65] proposed that brood parasitism evolved from an ancestor with parental care. Since
131 then, three key pathways have been suggested: by evolving directly from parental care [66],
132 via conspecific brood parasitism as a ‘stepping stone’ [67,68], or via cooperative breeding as a
133 precursor [69]. Ecological conditions and changes in life-history traits are also likely to have
134 influenced the transition [69,70]. However, these hypotheses remain surprisingly difficult to
135 test and understanding the origins of avian brood parasitism remains one of the most
136 fundamental open questions [42]. Molecular methods in the early 2000s produced a more
137 robust phylogeny for brood parasitic birds than had been available previously [71], and
138 facilitated more accurate estimates of when, and in which families, brood parasitism evolved
139 (e.g. [17,71–73]. Adopting this new phylogeny [71], however, changed conclusions of analyses
140 into the life-history conditions that surrounded the evolutionary transitions to brood parasitism:
141 for example, first comparative analyses using Aragón’s [74] and Johnson’s [75] molecular
142 phylogenies suggested that brood parasites evolved to reduce the costs of reproduction
143 associated with migration and a change in diet [76], whereas a similar analysis using the new
144 phylogeny with a different topology and branch lengths suggested migration evolved after
145 parasitism [75,77]. With only seven independent evolutionary origins of brood parasitism, can
146 we expect to ever know the conditions that facilitate it?

147

148 To understand the challenges of using the phylogenetic comparative method (PCM) to infer
149 the ecological drivers of evolution, it is useful to consider recent development in the field more
150 broadly. Genomic data and the methods used to build phylogenies and conduct comparative
151 analyses have become increasingly more accessible [78,79], but also more complex [80–83].
152 Essentially, increasing data (in terms of both loci sequenced and taxa sampled) has led to more
153 cases of discordance between gene trees and the species tree (e.g. because of introgression and
154 incomplete lineage sorting, [52,84,85]. This in turn has led to a conceptual change ([86] where
155 the initial assumption that the "true" species tree could be reconstructed from the average signal of
156 multiple genes has gradually been replaced by evolutionary models that incorporate various
157 mechanisms now known to cause and sort genetic variation differently across the genome (e.g.
158 varying mutation rates, recurrent hybridisation, gene flow, drift). The revolution of
159 phylogenetics to phylogenomics has also meant that applying PCMs to answer eco-

160 evolutionary questions is now more vulnerable to inferential errors as model assumptions are
161 not always considered [81,87]. Multiple evolutionary pathways can lead to the same outcome
162 (e.g. [88] and although correlation from PCM can reject or give support to certain hypotheses,
163 it cannot prove causality. For example, a PCM study by [89] suggested that brood parasitism
164 promotes cooperative breeding in hosts, whereas later work concludes the opposite - brood
165 parasites are attracted by cooperatively breeding hosts [90]. Therefore, it is necessary to
166 formulate accurate hypotheses that enable robust tests [91], and preferably use methods that
167 can compare multiple different hypotheses simultaneously (e.g. [92,93]. Using such an
168 approach, Griesser et al. [94] demonstrated that a 2-step process including both family living
169 and subsequent variable environmental conditions best explains the evolutionary origins of
170 cooperative breeding.

171
172 Even the most sophisticated comparative methods are, however, dependent on the quality of
173 data in inferring the origins of brood parasitism (e.g. [42,78]. Existing brood parasite
174 phylogenies could benefit from additional sampling of missing species and molecular markers.
175 For example, [71] used only a small fraction of the mitogenome and no nuclear markers,
176 leaving some of the older branching events and relationships among subfamilies unresolved.
177 Among brood parasites and hosts, ecological data has traditionally been highly biased towards
178 the temperate regions (e.g. [32,91], leaving replicated data on reproductive mode and other life
179 history traits of brood parasite and host taxa in the tropics lagging behind (although this is
180 improving, see [95]).

181
182 While we do not yet have data for a full comparative analysis using the PCM across all birds,
183 determining the conditions that facilitated the origins of brood parasitism is within our grasp if
184 we split the question into smaller steps (e.g. [96]. First, PCM can be used to infer trait evolution
185 linked with parasitism within lineages where all extant taxa have been sampled [97], dense
186 sampling). Second, sequencing and assembling whole genomes of non-model species opens up
187 the possibility to use comparative genomics and transcriptomics to investigate the origins of
188 avian brood parasitism and bypass the issue of only seven independent origins. For example,
189 Jackson et al. [98] were able to pinpoint evolutionary innovations associated with becoming a
190 parasite by comparing gene networks across genomes of three species: kinetoplastids,
191 protozoan parasites and their free-living relatives. When more functional annotations of genes
192 involved in parental care become available, it also becomes possible to track the evolution of
193 brood parasitism back in time using extant species. Given this framework, genomic

194 comparisons of closely related brood parasites and their non-parasitic sister species or even
195 within facultative brood parasite species (e.g. [99]) could also prove informative. While the
196 specific functions and regulatory pathways of genes associated with behavioural traits in higher
197 organisms remain poorly known ([Box 1](#)), the first steps towards understanding the molecular
198 basis of behavioural traits related to avian brood parasitism have already been taken. Lynch et
199 al. [100] used transcriptomics to compare gene expression between parasitic and non-parasitic
200 blackbird (Icterid) species, focusing on the preoptic area in the brain previously known to affect
201 maternal behaviour. In species without maternal care, gene expression patterns remained more
202 juvenile-like (i.e. neotenic) in adults. More detailed studies of different functional networks of
203 genes related to brood parasitism and ecological conditions within and between the
204 evolutionary lineages of extant brood parasites are thus likely to illuminate not only the
205 mechanisms, but also the ultimate causes behind becoming an obligate brood parasite.

206

207 2) When (and how) should hosts defend?

208

209 Given the fitness costs of raising a parasite, explaining why many host species lack defences
210 (e.g. almost 40% of avian brood parasite hosts tested do not remove foreign eggs from their
211 nest [38,101]) remains unresolved. In part, this is because it is challenging to tell from
212 behaviour alone whether a defence trait is yet to evolve (i.e. evolutionary lag, [102]) or is
213 present but there are insufficient cues to elicit its expression (i.e. cryptic plasticity, [103]).
214 Many brood parasites make use of behavioural and phenotypic tactics ([figure 2](#)) that increase
215 the costs of host recognition errors to achieve effective parasitism [38,101]. This then, in
216 theory, has selected for plasticity, with some hosts becoming more likely to defend with
217 repeated experience of parasitism (e.g. magpies become more likely to reject cuckoo eggs as
218 they age [104], or in response to environmental cues about local risk (e.g. reed warblers adjust
219 mobbing of adult cuckoos and rejection of eggs based on increased mobbing behaviour of
220 neighbours [105–107]). However, an alternative explanation for a lack of defences is that,
221 instead of preventing successful parasitism events (i.e. ‘resistance’), hosts may mitigate the
222 fitness costs of raising a parasite (‘tolerance’, [108,109]). Tolerance could, for example, include
223 hosts altering their behaviour or adjusting life history strategies by reducing investment in
224 current clutch size to save resources for later broods [110,111]. In other words, hosts without
225 apparent ‘defences’ may actually still be defending their fitness from a brood parasite [112].
226 Testing whether defences (or tolerance) have not yet evolved (or are cryptic) has, however,

227 proven challenging [110,111,113]; This is because of two major stumbling blocks: (i) it is
228 difficult to measure lifetime reproductive success and fitness of hosts (but see [104], and (ii)
229 field methods to test for plasticity in host defences are still rarely employed [114] and require
230 careful efforts to rule out cryptic defence traits. Could genomic tools be used to resolve whether
231 the lack of defences in hosts reflects a missing adaptation (or loss of trait due to relaxed
232 selection) or an alternative strategy (i.e. tolerance), given the inherent plasticity of behaviour?

233

234 One potential way to overcome the limitations of tracking individual fitness through
235 time in the field could be to use population genetic theory to ask if observed genomic patterns
236 fit one of the evolutionary scenarios (i.e. missing adaptation vs. tolerance). For example, if we
237 have both behavioural and genomic data from one host population from before and after
238 invasion by cuckoos, we could use simulations to compare allele frequency changes to a neutral
239 model without selection (i.e. using coalescent theory [Glossary](#), [25]). Such changes can then
240 be compared to the output of flexible evolutionary simulation frameworks (SLiM, [115];
241 Nemo, [116]), which would take into account key evolutionary parameters inferred from the
242 populations of interest. In addition, analysing genomic data from behaviourally tested rejector
243 and acceptor individuals from e.g. a recently parasitised population (where selection for
244 rejection is strong) and contrasting results to those of evolutionary simulations could help better
245 quantify the fitness effects of rejection behaviour. However, study systems sampled in at least
246 two time points with a known population history can be difficult to find in the wild (but see
247 [117] for some particularly amenable study systems under range shifts).

248

249 Given the difficulties of “finding a gene for behaviour” ([Box 1](#)), it is easier, and perhaps
250 also more meaningful, to study the genomic architecture of the trait (number and genomic
251 distribution of loci involved, along with their functional interactions) and its evolution in a
252 phylogenetic context (as in [118]). For example, by focusing on host species with well-
253 developed field methods to detect plastic trait expression (e.g. reed warblers), genomes could
254 be compared between rejector and non-rejector individuals to identify candidate genomic
255 regions associated with egg rejection defences. This could be done with association analyses,
256 which measure the correlations between genotypes and either phenotypic traits (Genome-Wide
257 Association Study, GWAS) or environmental variables (Gene-Environment Association, GEA,
258 [119]). This could also be conducted using FST scans which identify genomic regions of high
259 differentiation between populations from divergent environments or contrasting phenotypes

260 [120]. We could also generate high-quality reference genomes and compare chromosome-level
261 synteny across host species to identify e.g. candidate inversions associated with behavioural
262 traits [118]. This approach would work best with species that show little to no plasticity in their
263 defences (e.g. Cowbird hosts [121,122]), but, together with the intraspecific approach, results
264 could be used in future to develop genomic markers to detect defences in species where either
265 behavioural data is unknown or difficult to obtain, or where they may be cryptic. Identifying
266 the genomic regions underlying defences would also improve our understanding of trait
267 evolution [56]. For example, if egg rejection behaviour is associated with an inversion, then it
268 may be difficult to evolve as it requires large changes in the genome. However, it could arise
269 via introgression from other populations (see Question 3) or during speciation [123]. Discovery
270 of inversions have revolutionised our understanding of the evolution and maintenance of
271 polymorphisms in social behaviour (e.g. fire ant [124]) and ruff mating strategies (e.g. [125]).
272 However, all these approaches would benefit from more phenotypic data, collected across a
273 broader range of species, that takes plasticity into account.

274

275 3) Will coevolution persist across time and space with ecological 276 change?

277

278 As discussed in Question 2, all potential host species should in theory evolve defences and
279 render parasitism untenable given sufficient time (also see [38]). How is it possible then, that
280 brood parasites have persisted for millions of years (Question 1)? Furthermore, and perhaps
281 most importantly, can we predict what will happen next to hosts and brood parasites, given
282 rapid environmental change? The Geographic mosaic of coevolution theory (GMT, [126,127])
283 provides a compelling framework to answer these questions. First, it explains how antagonistic
284 coevolution can continue for long periods: local environmental variation, population dynamics
285 and demographics, gene flow, mutation and drift combine to produce mosaics of reciprocal
286 selection ('hot spots') and non-reciprocal selection ('cold spots') in time and space ([figure 3](#)).
287 The key cold spot that allows parasites to persist is where selection on hosts is sufficiently
288 relaxed that they lose their defences. This then allows parasites to eventually invade (although
289 some work suggests that it is variation in the parasite's virulence that determines long term
290 success [128]). Second, insight into the component parts of GMT can better facilitate predictions
291 about adaptive potential and interacting species' resilience to rapid environmental change
292 [129–132]. However, there have been few attempts to test GMT with avian brood parasites and

293 their hosts [114], despite them being a putative example in the theory's seminal publications
294 [126,127]. Behavioural experiments have revealed spatial correlations between host and
295 parasite traits (e.g. reed warblers vs. common cuckoo: [133–136]; magpies vs. great spotted
296 cuckoo: [137]; prinias and parrotbills vs. common cuckoo: [138,139], providing evidence for
297 hot spots as well as cold spots where parasitism is absent and defences vary, but we lack
298 quantitative estimates of the strength of selection and trait remixing at the genomic level. These
299 estimates are essential [140,141] to explain how coevolutionary interactions persist in time
300 [142] There have been some attempts to quantify gene flow and local allele frequencies of a
301 putative candidate marker for egg rejection in magpies [47], but this is where studies on brood
302 parasitism, and tests of GMT for behavioural coevolution more broadly, have hit a stumbling
303 block.

304

305 How can we quantify selection mosaics and trait remixing in avian brood parasite-host systems,
306 given that the heritable mechanisms of coevolved traits remain largely unknown
307 (Introduction)? Here again, combining empirical data with new molecular methods could offer
308 ways forward. Landscape genomics (table 1) uses genomic and environmental data collected
309 across a species distribution, or along an environmental gradient of interest. Neutral markers
310 are used to infer the underlying spatial and genetic population structure, and any remaining
311 markers associated with the environmental variable (e.g. current parasitism rate) are evidence
312 of selection. However, this method has rarely been used to test for associations with biotic
313 variables (e.g. [143]). The space-for-time substitution inherent to landscape genomics [144]
314 could also help solve a major issue: detecting selection usually requires long-term data [145].
315 However, behavioural studies are rarely replicated in time and space, although recent invasions
316 of brood parasites to new areas and range edge populations provide unique opportunities to
317 observe coevolution in action [117]. Studying spatial variation in egg polymorphism [146] and
318 egg rejection [117] has been suggested to overcome the lack of long-term behavioural datasets,
319 but there have thus far been no attempts to evaluate the genomic consequences of
320 environmentally varying brood parasitism risk in avian hosts.

321

322 Complementarily, we could use comparative phylogenetics (Question 1) between different
323 hosts and their parasites to look at host-parasite interactions at varying stages of coevolution
324 [117]. For example, parasitic lineages of *Viduidae* (parasitic whydahs) and *Anomalospiza*

325 (cuckoo finches), common cuckoos, and cowbirds have been estimated to have existed for 13,
326 6-8 and 3-4 mya, respectively [147–150], providing considerable variation in the duration of
327 potential coevolutionary interactions. As more molecular data and sophisticated analysis
328 methods become available, this variation could be used to resolve whether brood parasites and
329 hosts evolve at different rates, how brood parasites diverge, and the timing of host
330 specialisation [149–151].

331

332 With these landscape and comparative genomic approaches, it becomes possible to directly
333 measure levels of selection, gene flow and trait remixing. In systems where we have detailed
334 knowledge of host and parasite behaviour, demography, and genomics, it could also be possible
335 to move beyond the phenotypic gambit. For example, genome-wide markers have been
336 associated with climate adaptation in yellow warblers, and used to predict future vulnerability
337 to climate change [27]. Yellow warblers, however, are also a common host of the brown-
338 headed cowbird and the expression of host defences and levels of parasitism vary
339 geographically [152]. These data could be used to zoom in and look for the molecular
340 mechanisms underlying behavioural coevolutionary adaptations ([Box 1](#)). Molecular resources
341 are also becoming available for the reed warbler [64] to complement the wealth of existing
342 knowledge on geographic variation in defence behaviours [136]. Studying such avian brood
343 parasite and host systems would complement current molecular studies on biotic selection, as
344 examples of behavioural coevolution are few (e.g. ants: [128]), and most studies of the genomic
345 changes associated with defence traits in antagonistic coevolution come from systems in vivo
346 (e.g. [153]). However, accurate predictions of future coevolutionary trajectories requires
347 further development of models that can disentangle genomic patterns caused by processes other
348 than selection (i.e. migration in the molecular sense, mutation, recombination and drift). This
349 is a major goal in population genomics [22,154], and methods are rapidly developing (e.g.
350 [155]). In summary, analysing genomic data along with behavioural data in a brood parasite-
351 host system in the wild could not only test the theory of a geographic mosaic of coevolution
352 using direct genetic evidence but also help study three avenues of inquiry: speciation of hosts
353 and parasites, persistence of antagonistic coevolutionary interactions in time and maintaining
354 resilience to rapid ecological change.

355 Concluding remarks

356 Here we have shown that investing in analysing genomic data along with behavioural data in a
357 brood parasite-host system in the wild could lead to advances in understanding the evolutionary
358 origins of behavioural strategies, the fitness outcomes of plastic trait expression, the persistence
359 of antagonistic coevolutionary interactions in time, and how resilience to rapid ecological
360 change may be maintained. These themes are of broad interest beyond avian brood parasitism,
361 and show that genomic tools can be used to find answers to more than mechanistic questions.
362 By applying genomic comparisons at different levels ranging from within individuals to
363 between populations and species, we can also address how behaviour develops or changes
364 through ontogeny, its adaptive value, and its evolutionary past. In other words, genomic tools
365 can be integrated with behaviour to find answers to all of Tinbergen's 'four questions' (see
366 Box 2/Table 1, [Outstanding questions](#)).

367

368 Although we focussed our review on how genomic tools could be used to answer questions
369 about function, evolutionary history, and plasticity, having a mechanistic understanding of trait
370 heritability ([Box 1](#)) would nevertheless be beneficial to further advance many of the approaches
371 we've described here. For example, knowing the "genes for egg rejection" would allow us to
372 track its evolutionary history across taxa (Question 1), make it possible to determine the
373 'rejector' status directly from the genotype without behavioural testing (i.e. 'reverse ecology',
374 [156], disentangle the heritable and plastic components of trait expression (Question 2), collect
375 direct evidence of selection (Questions 2 & 3), test the theory of a geographic mosaic
376 maintaining coevolutionary trajectories in time (Question 3) and even predict future
377 adaptations [128]. Ideally, the underlying genomic mechanisms of plastic traits would be
378 studied by measuring levels of gene expression during a behavioural response. This popular
379 method can reveal the genes and regulatory mechanisms underlying differences in egg rejection
380 behaviour even if the responses are fully plastic and produced e.g. by post-translational
381 mechanisms from the same genotype, but are currently limited by uncertainty over which tissue
382 (and when) to sample ([Box 1](#)). However, field methods could be developed to experimentally
383 induce plasticity to the maximum and rule out cryptic defence traits, and emerging studies on
384 host hormone levels could guide which tissues to target, as hormone receptor expression is a
385 crucial step in the molecular pathways of behavioural responses [157]. Gene expression
386 experiments are, however, usually lethal which is not only problematic ethically but also means
387 we are unable to continue behavioural measurements to probe individual variation and

388 plasticity. Developing and validating methods using blood samples may provide a useful
389 alternative (e.g. [158]) although the crucial step of functional validation of (plastic) candidate
390 genes (using e.g. CRISPR, see Table 1) remains unlikely for avian brood parasites and their
391 hosts (see [Box 1](#)).

392

393 Throughout this article we have argued why integrating genomics into behavioural ecology
394 could be beneficial, but behavioural ecologists could also help resolve several outstanding
395 issues in genomics. For example, genomics has been criticised for being data-driven rather than
396 led by hypotheses [159,160], and increasingly reliant on searching for correlations rather than
397 experiments to test causation [161]. At the same time, adopting a Tinbergian approach from
398 behavioural ecology could help move forwards as it integrates mechanisms with development,
399 function and evolutionary history to test hypotheses (see [162] for discussion of how this could
400 revolutionise many fields). Similarly, variation in behaviour is often perceived to be more
401 difficult to measure than physical traits, but the underlying heritability of both is intertwined
402 [163,164]. Behavioural ecologists could bring a deep understanding of how and what to
403 measure, as well as the ecology underlying the trait in question [30], to frame hypotheses
404 appropriately. We now have a plethora of sequencing techniques and massive datasets
405 becoming available. Utilising these data will require appropriate analysis methods that are
406 carefully designed to address questions about polygenic traits with inherent plasticity,
407 something best achieved by forming collaborative teams ([Box 2](#)) that make use of our wide
408 range of complementary skill sets.

409

410 Outstanding questions

- 411 ● Examples of other questions regarding avian brood parasitism: Do rates of evolution
412 vary between avian brood parasites and hosts? What are the molecular mechanisms
413 that facilitate rapid changes in egg coloration during evolving host-parasite arms
414 races? How is egg or plumage coloration polymorphism maintained?
- 415 ● What platforms are needed to best bring behavioural ecologists and molecular
416 ecologists together to better integrate Tinbergens' four questions?
- 417 ● If we find candidate genes for behaviours, can field-friendly methods be developed to
418 experimentally test gene function?

- 419 ● When does behavioural plasticity facilitate or hinder adaptation? This will require
420 deeper integration of behavioural experiments with epigenomics.
- 421 ● Can a richer understanding of behavioural interactions improve heritability estimates
422 in studies of genetic and non-genetic inheritance (or indirect genetic effects)?
- 423 ● Do new behavioural adaptations occur from selection acting on standing genetic
424 variation, introgression or new mutations?

425

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438

439

440 Glossary

441

442 *Bioinformatics* - Science of collecting and analysing biological big data (e.g., nucleic sequences,
443 genomes, gene / protein expression, biological networks, single cell data).

444

445 *Coalescent theory* - A population genetics model that allows reconstructing genealogies of all alleles
446 of a gene in a sampled population. Coalescent theory treats genealogies as random providing null
447 models for testing causes of genetic variation.

448

449 *DNA* - The deoxyribonucleic acid (DNA) is a macromolecule consisting of two chains forming a
450 double helix and including four nucleotides: adenine, thymine, cytosine and guanine. The nucleotide
451 sequence contains heritable instructions for the development, function, growth and reproduction for
452 all living organisms (and some viruses).

453

454 *Epistasis* – Phenomenon where the effect of one mutation on a phenotype is dependent on the
455 presence (or absence) of additional mutations in at least one gene.

456 *Evolutionary simulations* - Simulations of realistic, individual-based, and genetically explicit
457 evolutionary scenarios using dedicated computer programs (e.g., nemo, SLiM). These programs are
458 highly flexible and allow simulating hundreds of generations of evolution in terms of population
459 genetics and/or life history trait (e.g., host-parasite relationships).

460

461 *Gene* - A philosophical concept of the unit of inheritance, which may often correspond to e.g. a
462 protein-coding DNA sequence. However, it is no longer clear where “a gene” begins or ends in the

463 genome as the mechanisms of reading DNA have turned out to be various. Analogous to a word
464 describing an ingredient if the genome is the cook-book.
465
466 *Genetics* - Study of genes, genetic variation and their heritable effects on organisms.
467
468 *Genome* - All genetic material of an organism, consisting of DNA (or RNA), and found in the nuclei
469 of (almost) every cell of a multicellular organism. In a sense, contains all heritable information needed
470 to build and operate the organism, i.e. the cook-book for an organism and its functions.
471
472 *Genomics* - Field of biology focusing on the structure, function, and evolution of genomes.
473
474 “*Next-generation genomics*” - A term referring to methods developed since 2005 and which aim at
475 reconstructing the nucleotide sequence (sequencing). “Next-generation genomics” (NGS) differ from
476 traditional Sanger sequencing in that billions of small DNA fragments can be sequenced
477 simultaneously (shotgun sequencing). More descriptive names include massively parallel sequencing
478 or high-throughput sequencing.
479
480 “*Third generation genomics*” - This term refers to methods sequencing very long molecules of DNA
481 (> 10,000 nucleotides) contiguously (such as PacBio [165]), which help assembling reference
482 genomes and identify structural variations such as inversions.
483
484 *Post-genomics* - Era starting after the completion of the Human Genome Project, where life sciences
485 progressed beyond a gene-centred view to better understand genome functions and its evolution.
486
487 ‘Omics’ - An informal term for several branches of science in biology aiming to characterize and
488 quantify pools of biological molecules that translate into organisms’ structures, functions and
489 dynamics. Examples of omics include genomics, metagenomics, transcriptomics, metabolomics and
490 proteomics or their integration in multiomics.
491
492 *Gene expression* - Process by which the information contained in the gene sequence is used to
493 synthesise a gene product (usually a protein). This can be measured for single genes or across the
494 genome (with RNA-seq, see [Table 1](#) for transcriptomics) from different tissues and/or conditions.
495
496 *Introgression* - Transfer of genetic material between genetically distinct populations (e.g., species)
497 through hybridization.
498
499 *Inversion* - Reversal of the orientation of a section of chromosome. Within the inversion,
500 recombination is suppressed between inverted and non-inverted alleles.
501
502 *Microsatellite* - Tract of DNA made of the repeat of a single motif, composed of few nucleotides. The
503 number of repeats is highly variable (polymorphic) between individuals, and for this reason
504 microsatellite loci have been used as genetic markers since the late eighties.
505
506 *Plasticity* - When the same genotype produces different phenotypes according to environmental
507 conditions.
508
509 *Pleiotropy* – When variation at one gene influences two or more different, seemingly unrelated traits.
510
511 *Phenotypic gambit* - an assumption that the genetic mechanisms underlying behavioural traits can be
512 ignored or modelled as simple biallelic alternative strategies. Instead, one can determine the expected
513 evolutionary dynamics of a population based on fitness of the different phenotypes.
514
515 *QTL* - A quantitative trait locus (QTL) is a section of DNA correlating with variation in a complex
516 trait (for which phenotypic variation is continuously distributed in natural populations, e.g. height).

515 QTLs are identified (or ‘mapped’) by measuring the statistical association between genetic markers
516 (such as SNPs, microsatellites) and a trait of interest, usually using crosses between lineages differing
517 in their average trait values.

518 *SNP* - A single nucleotide polymorphism (SNP) is a substitution of one base pair at a given position in
519 the genome. Analogous to a letter in a word that is the gene.

520

521 Box 1. Behavioural genomics in the wild

522 Many of the previous calls for behavioural ecologists to adopt genomic tools have focused on
523 using approaches from behavioural genomics [14–16,166]. Here, the goal is to uncover the
524 molecular mechanisms underlying behavioural traits and ideally this requires measuring levels
525 of gene expression during a behavioural response (within the whole organism, organs or single
526 cells; see transcriptomics in Table 1), determining which genes are up- or downregulated
527 compared to when the behaviour is not expressed, then experimentally testing causation by e.g.
528 gene knock-out experiments or crosses (gene editing in Table 1). Most of the methods
529 available, however, were originally developed using model organisms (e.g. humans,
530 *Drosophila*, zebra fish, laboratory mice and rats), and much of the success thus far has come
531 from studying organisms that lend themselves to highly controlled laboratory-based
532 experiments (e.g. great tits, [167]) or where lethal sampling of tissues has fewer ethical
533 concerns (e.g. Heliconius butterflies: [168]). This limits wide-scale adoption to study
534 behavioural traits in wild populations. For example:

535

536 **quantitative trait locus** (QTL) analyses designed to account for polygenic heritability require
537 large sample sizes of behaviourally-phenotyped and pedigreed individuals, coupled with long-
538 term monitoring, which can be difficult to obtain and require significant resources, including
539 long term funding, for organisms in the wild [18].

540

541 Top-down approaches such as **genome-wide association studies** (GWAS) are most effective
542 when traits are determined by few genes or ‘supergenes’ and associations are not masked by
543 environmentally induced plasticity ([Glossary](#), see [Table 1](#) for examples). However, large-effect
544 loci and simple genetic architectures only occasionally explain phenotypic variation (e.g. [18])
545 and identifying small effect genes contributing to behavioural traits has proven challenging
546 even in humans, despite hundreds of thousands of genomes sequenced [169]. In addition,
547 regardless of the genetic architecture, environmental variation may only affect the presence of
548 alleles in one setting (i.e. conditional neutrality, [170]), or produce an opposing gene expression

549 pattern in two different environments (i.e. plasticity), meaning that the different phenotypes
550 cancel out the association. GWAS methods are however being developed to account for
551 confounding variation (e.g. naturalGWAS, [171] and RepeatABLE [172].

552

553 Bottom-up **gene expression studies** depend on the researchers' ability to induce and measure
554 relevant and consistent behavioural responses (including the control or reference behaviour),
555 and determine the correct timing and location to sample tissue where the genes are expected to
556 be expressed (which can be measured within the whole organism, organs or single cells). This
557 is especially problematic when studying behaviour in the wild as there is substantial uncertainty
558 regarding where and when to sample expression of genes and behaviour [173]) and most tissues
559 require lethal sampling.

560

561 Furthermore, the lack of relevant **functional annotation** of genes and knowledge of
562 **regulatory gene networks** is a major hindrance for any genome-wide or gene expression
563 studies in the wild [15,30], although available annotations for markers expressed (or
564 underexpressed) are increasing, e.g. social behaviour in quail [174].

565

566 Finally, **gene editing approaches** using methods such as CRISPR-cas9 ([Table 1](#)) to
567 experimentally test associations are not feasible (or ethical) with most wild vertebrate study
568 organisms [175] and are still largely a blunt tool to probe complex traits.

569

570 More broadly, there is ongoing debate as to whether searching for a 'gene (or genes) for
571 behaviour' is worthwhile (e.g. [33,169]). Detailed studies on the 3D-structures of (human)
572 genomes and epigenomics are revealing that phenotypic traits are often determined by complex
573 regulatory pathways, affecting the timing of expression in networks of tens to hundreds of
574 interacting genes (e.g. epistasis and pleiotropy, [Glossary](#)), and even the formation of
575 supergenes is regulated by several genes associated with hormones [176]. These developments
576 have led to a paradigm shift in genomics from focusing on gene sequences, to understanding
577 the regulatory and evolutionary mechanisms occurring at the molecular level [3,6], see e.g.
578 [177] for insights from single-cell genomics). In this review, we therefore outline approaches
579 and methods to be used with wild organisms without the need to 'find a gene for behaviour' to
580 advance the field of behavioural ecology with genomic data and the most appropriate tools that
581 are readily available.

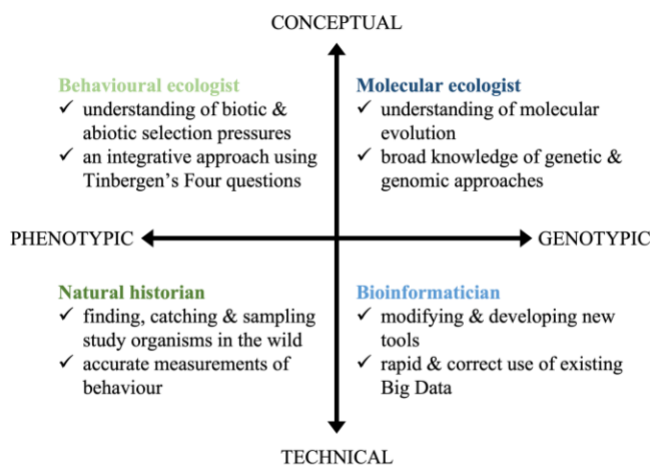
582

583 **Box 2. Forming collaborative teams to take the field forward?**

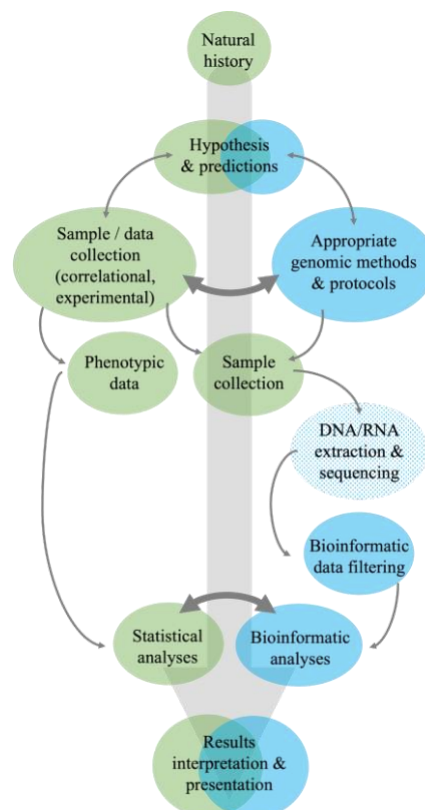
584 Behavioural ecology was built on the foundation of Tinbergen’s four questions [20]: we can
 585 only understand a behavioural trait by investigating its underlying mechanisms, how the trait
 586 develops, its evolutionary history, and its function (i.e. effect on fitness). Tinbergen stressed
 587 that answers to each question were complementary rather than mutually exclusive, and
 588 behavioural ecology has since grown into one of the most integrative fields in biological
 589 sciences [178]. While genomic methods could become a useful part of our toolkit to answer
 590 aspects of each question, our goal is not to advocate for all behavioural ecologists to become
 591 experts in genomics. Neither should molecular ecologists necessarily all become experts in
 592 behaviour. Rather, we should form collaborative teams that make use of our wide range of
 593 complementary skill sets ([Box 2 figure 1](#)): the four specialists represent the different levels of
 594 inquiry from conceptual question framing to technical problem solving and move between
 595 phenotypic and genotypic approaches. Naturally the number of people does not need to be
 596 four - many scientists may sit closer to the centre on both axes and thus bridge the gap
 597 between solely phenotypic or genotypic approaches.

598

(a)



(b)



599

600 Box 2 figure 1. (a) Composition of an ideal collaborative team to bridge the gap between
601 molecular and behavioural ecology. (b) A hypothetical workflow of behavioural ecologists
602 and natural historians (green) working with molecular ecologists and bioinformaticians
603 (blue), from project conception to completion (large grey arrow). Smaller arrows indicate
604 each step with their width highlighting essential points in the collaboration. Note that
605 DNA/RNA extraction and sequencing (shown as stippled) is likely to be outsourced.

606

607 To illustrate the potential benefits of interdisciplinary collaboration here we present a
608 hypothetical workflow ([Box2, figure 1b](#)) from planning, data collection and analysis to
609 reporting. Behavioural ecologists ask *hypothesis-driven* questions based on detailed
610 knowledge of *the natural history of the study organism* to account for different selection
611 pressures and confounding factors. However, jumping to the genomic era will require
612 detailed knowledge of the *available methods and the theory* behind them to enable efficient
613 communication between fields. Collecting *accurate behavioural data* and analysing it in a
614 meaningful way, or even finding and sampling *enough individuals* in their natural habitats
615 requires field skills. DNA extraction and sequencing can often be outsourced, yet the
616 subsequent preprocessing of the massive genomic data and *choosing appropriate methods* to
617 draw biologically meaningful inference from it requires knowledge of *bioinformatics*.
618 Finally, collaborative discussions throughout the process will aid the team in reporting their
619 results and conclusions in a precise but understandable way to the benefit of wider audiences
620 than may be currently interested in behavioural evolution.

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633 Table 1. Genomic tools to study behaviour.

634

635 These include common methods and -omics approaches already utilised in behavioural
 636 ecology (see figure 1). Example studies include behavioural traits but are not necessarily
 637 from behavioural ecology. Suggestions for further reading provide broad overviews.

638

Tool	Definition	Goal	Example studies	Further reading
Genotyping	Determining differences in individual genotypes	Identifying DNA markers for defining biological populations, parentage, breeding or tracking disease	[179] and [180] for great tit HapMap projects.	[181]
Quantitative genomics	Statistical analysis of quantitative trait loci (QTL)	Inferring the genetic basis and heritability of polygenic traits	[182] found a simple oligogenetic basis for mate recognition in <i>Heliconius</i> butterflies	[183]
Microbiomics	Study of the microbiome, which is the community of micro-organisms (e.g. bacteria, protozoa, fungi and viruses) inside a given habitat (e.g. body part, organism or environmental sample)	Understanding the factors causing variation in microbiomes and how the microbiome interacts with its environment	[184] for burying beetles regulating carcass microbiota for their offspring, [185] for factors affecting wild bird microbiota.	[186]
Phylogenomics	Study of evolutionary relationships using phylogenetic inference drawn from genomic data	Testing evolutionary hypotheses either by sequence-based methods or whole-genome features	[94] for evolution of cooperative breeding in birds. For other examples see Question 1	[187]

Comparative genomics	Comparisons of the genomic features of organisms	Understanding heritability and genomic evolution across organisms	[188] to investigate genes related to tool use in corvids	[189]
Transcriptomics	Study of all ribonucleic acid RNA in a cell	Detecting genes and defining regulatory pathways underpinning trait expression	[190] aggressive host responses to cowbirds	[177,191,192]
Epigenomics	Study of epigenetic processes in the genome, such as methylation and altered expression rates.	Understanding variation in behaviour that is not determined by genotype	[193] parent exposure to predation alters offspring mating behaviour in three-spined sticklebacks.	[4] for discussion on evolutionary implications and twin studies.
GWAS	Genome-wide studies of associations between genetic markers or functional gene networks and a phenotypic trait	Detecting relevant loci to study molecular mechanisms underlying phenotypes of interest	Supergene determining behavioural lekking strategies in the ruff [194] and [195] for resistance genes in <i>Daphia</i> against a bacterial pathogen	[196]
Gene editing	Manipulation of the genome by deleting, inserting, or replacing a gene sequence	Producing transgenic individuals to test gene functions	[197] used RNAi to knock-out genes in earwigs to study parental care. [198] used CRISPR/CAS9 to knock-out genes associated with social behaviour in raider ants.	[199–201]
Population genomics	Study of evolutionary processes at the	Understanding populations, microevolution,	[202] for population structure across	[203], [204]

	population level with genomic concepts and technologies	demography and phylogenetic history	elevations in Mountain chickadees	
Landscape genomics	Study of environmental variables associated with genetic adaptation to those variables	Demonstrating natural selection, studying spatial variation in adaptation	[205] used Gene-Environment Association analyses (GEA) to detect signs of selection during a range expansion in a damselfly, and Question 3	[144]

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Figure 1.

The use of genomic (i.e. next-generation and third-generation methods using high-throughput sequencing) and genetic (any non-genomic method utilising DNA or RNA) tools in behavioural ecology using (*a*, *b*) abstracts from International Society for Behavioural Ecology (ISBE) conference talks. Despite a dramatic reduction in sequencing costs (dashed line), genomic tools are used rarely in behavioural ecology (*a*), although a range of methods and approaches are applied across a wide range of topics (*b*). The titles and abstracts of accepted talks in ISBE conferences between 2006 and 2022 were checked manually from abstract booklets for mentioning usage of “genomic methods” (i.e. next generation, high throughput, deep or RAD sequencing, *omics*, transcriptom* or RNA-sequencing). Talks mentioning usage of molecular methods (or DNA or RNA) that were not genomic were counted as using “genetic methods”. Only regular talks with abstracts in the programs were included, leaving out plenaries, keynote speakers and cancelled talks from ISBE2022 from the total number of accepted talks. Only articles clearly stating the genomic methods used were included to make figure 1b. The included ISBE talks with abstracts are listed in the Supplementary Table 1.



Figure 2.

Examples of the arms races between avian brood parasites and their hosts. Row 1: virulence (a) low - bronzed cowbird with Bewick wren host chicks, (b) high - common cuckoo removing reed warbler's egg (c) mutualism - great spotted cuckoo and crow host; Row 2: defences (d) mobbing, (e) egg rejection, (f) chick rejection; Row 3: mimicry (g) hawk mimicry by adult *Cuculus* cuckoos, (h) *Prinia* egg mimicry by cuckoo finch, (i) host chick (left) mimicry by *Chalcites* cuckoo species (right).

Image credits:

(A) Rolf Nussbaumer / Alamy Stock Photo, (B) Richard Nicoll, (C) Vittorio Baglione

(D) Alan McFadyen / Scottishphotographyhides.com, (E) Oldrich Mikulica

(F) Alfredo Attisano, (G) left panel: Frans Lemmens/Alamy Stock Photo; right panel, Mike Lane/Alamy Stock

Photo, (H) Claire Spottiswoode, (I) Naomi Langmore

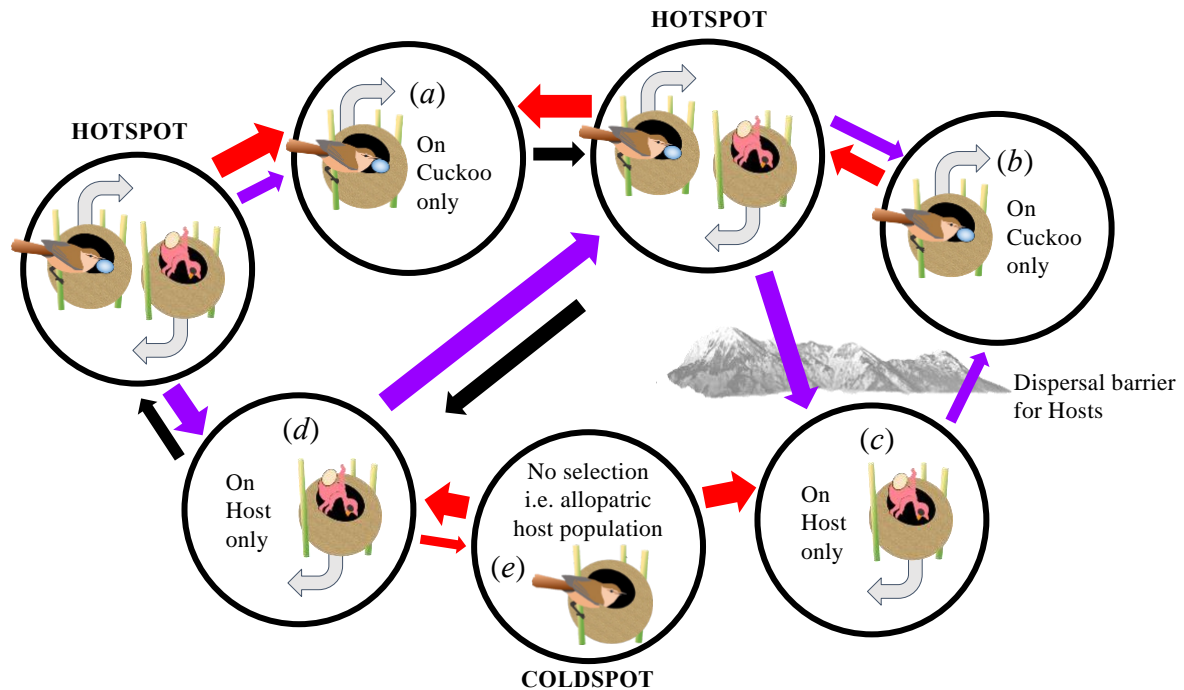


Figure 3.

A hypothetical geographic mosaic of coevolution between a brood parasitic cuckoo and a warbler host (key components in bold text, modified from [127]): Circles represent locations where **selection varies** and arrows between circles describe the direction of **gene flow** (red is host, purple is cuckoo, and black represents similar gene flow of both). Arrow thickness indicates **trait remixing** essential for maintaining genetic variation: if gene flow is absent then local fixation of alleles increases, whereas high levels of gene flow cancel out local adaptation. Reciprocal selection occurs in ‘hotspots’ and non-reciprocal selection occurs in ‘coldspots’; these vary according to local population dynamics e.g. (a) strong gene flow of experienced hosts or (b) locally fixed host defences exert stronger selection on local cuckoos, or (c) a recently invaded host population lacks defences or (d) ‘spill over’ of cuckoos from another host species exerts stronger selection on hosts. Coldspots also arise when (e) hosts and cuckoos do not co-occur (i.e. relaxed selection). Behaviour could influence each component: local environmental conditions determine the relative fitness of expressing or retaining behavioural defences (especially when plastic) when parasitism is low (i.e. strength of selection), movement of behavioural phenotypes is unlikely to be homogenous (i.e. affecting direction and specific genotypes during gene flow), and both defences and population dynamics can depend on the behaviour of others’ phenotypes/genotypes (i.e. indirect genetic effects).

Supplementary material

Table S1

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