

Understanding and applying biological resilience, from genes to ecosystems

Authors and affiliations:

* corresponding author (rose.thorogood@helsinki.fi)

§ equal contribution

All other authors listed alphabetically

Rose Thorogood*^{§1,2,3}, Ville Mustonen^{§2,4,5,6}, Alexandre Aleixo⁷, Pedro J. Aphalo^{2,8}, Fred O. Asiegbu^{8,9}, Mar Cabeza^{2,3}, Johannes Cairns^{2,4}, Ulrika Candolin², Pedro Cardoso⁷, Jussi T. Eronen^{3,10,11}, Maria Hällfors^{2,12}, Iiris Hovatta^{13,14,15}, Aino Juslén⁷, Andriy Kovalchuk^{9,16}, Jonna Kulmuni², Liisa Kuula¹³, Raisa Mäkipää¹⁷, Otso Ovaskainen^{2,18}, Anu-Katriina Pesonen¹³, Craig R. Primmer^{2,6}, Marjo Saastamoinen^{1,2,12}, Alan H. Schulman^{6,8,17}, Leif Schulman^{7,20}, Giovanni Strona^{2,12}, Jarno Vanhatalo^{2,12,19}

1. HiLIFE Helsinki Institute of Life Science, University of Helsinki, Helsinki, Finland
2. Research Programme in Organismal & Evolutionary Biology, Faculty of Biological and Environmental Sciences, University of Helsinki, Helsinki, Finland
3. HELSUS Helsinki Institute of Sustainability Science, University of Helsinki, Helsinki, Finland
4. Department of Computer Science, Faculty of Science, University of Helsinki, Helsinki, Finland
5. Helsinki Institute for Information Technology, University of Helsinki, Helsinki, Finland
6. Institute of Biotechnology, HiLIFE Helsinki Institute for Life Science, University of Helsinki, Helsinki, Finland
7. LUOMUS Finnish Museum of Natural History, University of Helsinki, Helsinki, Finland
8. Viikki Plant Science Centre, University of Helsinki, Helsinki, Finland
9. Department of Forest Sciences, Faculty of Agriculture and Forestry, University of Helsinki, Helsinki, Finland

10. Research Programme in Ecosystems and Environment, Faculty of Biological and Environmental Sciences, University of Helsinki, Helsinki, Finland
11. BIOS Research Unit, Helsinki, Finland
12. Research Centre for Ecological Change, Faculty of Biological and Environmental Sciences, University of Helsinki, Finland
13. SleepWell Research Program, Faculty of Medicine, University of Helsinki, Helsinki, Finland
14. Department of Psychology and Logopedics, Faculty of Medicine, University of Helsinki, Helsinki, Finland
15. Neuroscience Center, HiLIFE Helsinki Institute for Life Science, University of Helsinki, Finland
16. VTT Technical Research Centre of Finland Ltd, Espoo, Finland
17. Natural Resources Institute Finland (Luke), Helsinki, Finland
18. Centre for Biodiversity Dynamics, Department of Biology, Norwegian University of Science and Technology, Trondheim, Norway
19. Department of Mathematics and Statistics, Faculty of Science, University of Helsinki, Helsinki, Finland
20. SYKE Finnish Environment Institute, Helsinki, Finland

1 **ABSTRACT**

2 The natural world is under unprecedented and accelerating pressure. Much work on understanding
3 resilience to local and global environmental change has, so far, focussed on ecosystems. However,
4 understanding a system's behaviour requires knowledge of its component parts and their interactions.
5 Here we present a framework for understanding 'biological resilience', or the processes that enable
6 components across biological levels, from genes to communities, to resist or recover from perturbations.
7 Although ecologists and evolutionary biologists have the tool-box to examine form and function, efforts to
8 integrate this knowledge across biological levels and take advantage of big data (e.g. ecological and
9 genomic) are only just beginning. We argue that combining eco-evolutionary knowledge with ecosystem-
10 level concepts of resilience will provide the mechanistic basis necessary to improve management of
11 human, natural and agricultural ecosystems for better resilience.

12

13 **KEYWORDS** Eco-evolutionary dynamics; evolutionary ecology; macroecology; ecology; ecological
14 resilience; applied management; forestry; crops; health

15

16 **IMPACT STATEMENT**

17 Resilience to environmental change will depend on ecological and evolutionary processes operating
18 across all biological levels of organisation, yet integrating this knowledge for application is only just
19 beginning.

20 **Introduction**

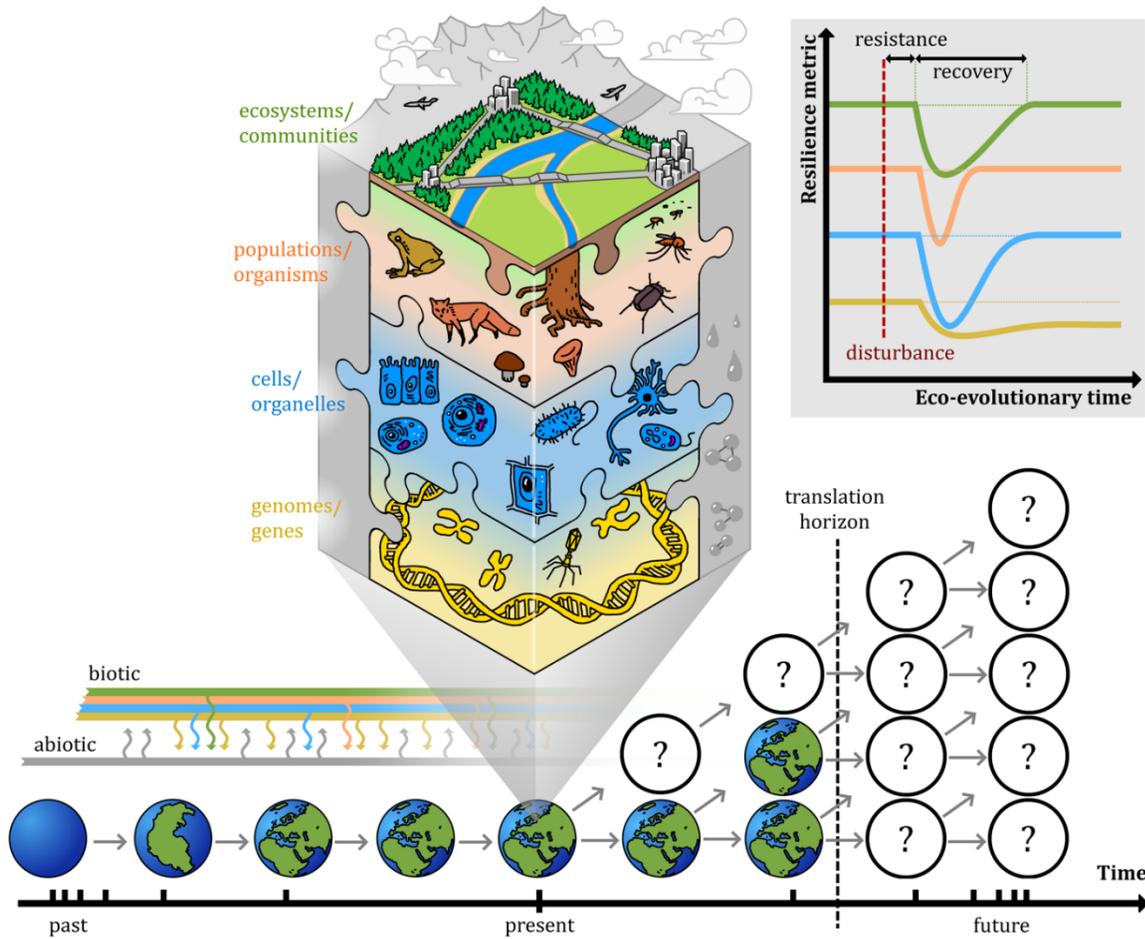
21 The Anthropocene is characterised by the pervasive impact of human activity on all aspects of life on
22 earth (Lewis & Maslin, 2015). Human-driven climate change and overexploitation of natural resources, as
23 well as increasing human population densities and urbanisation, are placing progressively larger areas
24 under human influence (Ellis, 2015), and conspicuous perturbations include increased and/or more
25 variable temperatures (and associated events such as droughts and fires), direct anthropogenic
26 alterations (e.g. pollution, land-use changes, habitat fragmentation), and invasive species. Even the
27 world's topology has changed, as global movement of individuals and goods erodes biogeographical
28 barriers (van Kleunen et al., 2015). These environmental changes put ecosystems under unprecedented
29 and accelerating pressures, inducing regime shifts (Scheffer, Carpenter, Foley, Folke, & Walker, 2001),
30 causing loss of ecosystem services (Foley, 2005), and even changing the course of evolution (Sullivan,
31 Bird, & Perry, 2017). There is therefore an urgent need to determine why some species, communities or
32 ecosystems decay while others persist or adapt (Sutherland et al., 2013), and then implement this
33 knowledge for improved management practices that can reverse or mitigate damage (Weise et al., 2020).

34
35 In ecology, 'resilience' has attracted great interest as a concept that describes how ecosystems recover
36 to an antecedent state following a disturbance ('engineering resilience'; Pimm, 1984), or absorb change
37 and resist large shifts in ecosystem function ('ecological resilience'; Holling, 1973). Ecological phenomena
38 are the result of a myriad and complex array of responses by different components across biological
39 levels, that may or may not interact. Therefore, resilience has typically been studied at the ecosystem
40 level (Capdevila et al., 2021) with current work recognising that the resilience of ecosystems is likely to
41 depend on both its ability to resist and recover following disturbance (Ingrisch & Bahn, 2018; Capdevila et
42 al., 2021). However, the mechanisms that determine whether an ecosystem resists, or what shapes the
43 trajectory of its recovery, remain largely unknown (Oliver et al., 2015; Capdevila et al., 2021). In part, this
44 may be because studying resilience at the level of the ecosystem reduces our power to identify how and
45 why resistance and/or recovery responses occur (Gladstone-Gallagher et al. 2019): understanding the
46 behaviour and interactions of a system's component parts is essential to understand and forecast ecology
47 (Levin, 1992). Furthermore, there has been much discussion as to how resilience can be measured, with

48 debate over what metrics are most relevant (e.g. Hodgson et al. 2015). Studying lower biological levels in
49 isolation might make these metrics easier to identify (e.g. population size, individual fecundity, genetic
50 diversity, Figure 1 inset) but it hampers detection of connections between seemingly isolated biological
51 events. How can we deal with this complexity to identify the critical components and indicators of
52 resistance and recovery?

53
54 Here we propose that this can be achieved by adopting a 'biological resilience' framework (Figure 1)
55 where we (1) test ecosystem-level resilience concepts (i.e. resistance and recovery responses to
56 perturbations; Pimm, 1984; Holling 1973) across lower levels of biological organisation and (2) harness
57 knowledge provided by the eco-evolutionary history of adaptation to past perturbations (Dakos et al.,
58 2019; Sgrò et al., 2011). In this way, biological resilience acknowledges that processes occurring within
59 and between components across biological levels, from genes to communities, shape how systems resist
60 or recover from perturbations. This framework stands out from recent calls to encourage analysis of
61 resilience across systems, scales, and biological levels (e.g. Gladstone-Galagher et al. 2019; Capdevila
62 et al. 2021) as we explicitly acknowledge the crucial role of eco-evolutionary history, and encourage
63 studies of resilience to dive deeper into uncovering the mechanisms and processes that afford resilience
64 within individuals (e.g. genetic diversity, cellular response) and how these scale up to affect populations,
65 communities and ecosystems. We first explore how the eco-evolutionary past provides context for present
66 and future resistance and recovery responses, and then discuss why it is necessary to consider how
67 abiotic and biotic perturbations can affect biological levels differently to detect mechanisms and
68 underlying processes. Next, we outline three testable hypotheses to kick-start research into resilience
69 across levels of biological organisation, from genes to cells, individuals, populations and communities.
70 Collecting and integrating large amounts of data about how every biological component responds to a
71 perturbation is often considered unrealistic. However, here we identify new opportunities emerging from
72 the ongoing infusion of big data into ecology and evolutionary biology and stress the need to combine
73 these data with experimental approaches. Finally, we discuss how considering resilience at the
74 appropriate biological level(s) will enable advances in translating research into practice.

75
FIGURE 1 HERE



76

77

78 **FIGURE 1.** Biological resilience, or the mechanisms and processes that enable components across
 79 biological levels to resist or recover from perturbations (inset), is mediated by connections within and
 80 among levels of life (simplified here to genes and genomes, cells and organelles, organisms and
 81 populations, communities and ecosystems). Taking a biological resilience approach requires integrating
 82 knowledge about how the present state (centre focus) has been shaped by ecological and evolutionary
 83 responses to biotic (depicted by multi-coloured lines and levels) and abiotic (grey) perturbations and
 84 selection pressures in the past (time represented by a log-scale). Knowledge of biological resilience can
 85 then enhance the translation horizon (vertical dashed line, close in time) by reducing uncertainty in
 86 prediction trajectories (grey arrows) and improving accuracy of forecast outcomes (denoted by question
 87 marks within circles). Note that the resistance and recovery trajectories of biological levels may differ in
 88 both amplitude and temporal scale.

89

90 **(Re)Placing resilience into an eco-evolutionary context**

91 When Holling introduced ecological resilience in his landmark paper (Holling, 1973), he briefly suggested
92 that a system's resilience is a product of its evolutionary history (1973:p.18). Most research conducted
93 since, however, has lacked an evolutionary perspective (McGill et al., 2019; Oliver et al., 2015) and
94 therefore much of the discussion, theory and examples of resilience in ecology lack a long time horizon.
95 Similarly, eco-evolutionary biologists rarely study resilience explicitly in terms of resistance or recovery
96 from a perturbation, nor do they consider how resilience might be conferred by processes that occur
97 within or across the biological levels that form the focus of their studies (Dakos et al., 2019). This
98 disconnect among fields may be because much of the work on resilience describes patterns at the
99 ecosystem level (Oliver et al., 2015), whereas studies of evolutionary processes rarely scale to complex
100 communities (Tylianakis & Maia, 2020). Indeed, focusing on how ecology and evolution shape patterns
101 and processes within individuals and populations has attracted criticism for being too narrow to address
102 large ecological problems (Carroll et al., 2014; McGill et al., 2019). Nevertheless, here we argue that
103 taking an eco-evolutionary perspective in understanding biological resilience can provide information from
104 the evolutionary past to improve our power to estimate both present and future states (Box 1). Rather
105 than 'reinventing the wheel', this approach to understanding resilience ties into calls to apply 'resilience-
106 thinking' from ecosystems to species (e.g Capdevila et al., 2021) and wider discussions in ecology and
107 evolution about the need to rediscover connections among the fields (e.g. McGill et al., 2019).

108

109 **Reading the past from the present state**

110 Past perturbations leave their mark on biological entities, creating ecological and evolutionary 'memories'
111 (Desai, 2009; Johnstone et al., 2016) that may influence responses to similar perturbations in future. At
112 the level of genes, evolutionary history is manifested in variation introduced by mutation and/or migration
113 (gene flow) as well as recombination (new combinations of genetic variation) that is filtered by natural
114 selection or fixed by random genetic drift. Some of these variants may provide an advantage against a
115 future perturbation, such as through acquired resistance against a parasite, pest or antibiotic encountered
116 in the past (Bartholomé et al., 2020). On the other hand, perturbations that result in severe population
117 bottlenecks can result in the loss of potentially beneficial variation and/or fixation of maladapted alleles,

118 and thus have negative effects on resilience (Donelson et al., 2019). Similarly, past selection that strongly
119 favoured specific alleles may also limit future resilience due to the loss of genetic variation required for
120 new adaptation to take place (e.g. Afrotropical butterfly experiencing climate change induced variation in
121 seasonality, Oostra, Saastamoinen, Zwaan & Wheat et al. 2018). Responses to past perturbations may
122 also leave heritable epigenetic marks on the genome without causing changes to DNA sequences. These
123 epigenetic 'stress memories' can affect how subsequent generations resist or recover from similar
124 disturbances by influencing the regulation of genes (e.g. Holeski, Jander, & Agrawal, 2012; Pazzaglia et
125 al., 2021), although the study of these processes beyond plants is only just beginning.

126
127 While genetic information should underpin the capability of an organism to respond (e.g. Waldvogel et al.
128 2020), there is now abundant evidence that single genotypes can generate different phenotypic
129 responses to the same environmental perturbation (i.e. plasticity). Phenotypically plastic responses can
130 be modified further depending on the composition, structure and spatial context of the perturbed
131 population or ecological community (Thebault & Fontaine, 2010), and plasticity that evolved in the past
132 can shape current distributions (Valladares et al., 2014). Behavioural modifications to a changing world
133 are a common form of plasticity (Wong & Candolin, 2014) and can enable individuals and populations to
134 resist fitness or demographic effects of perturbations. For example, when behavioural repertoires are
135 transmitted within or across species (i.e. via horizontal and/or vertical social learning), individuals can
136 acquire phenotypes that match environmental changes more quickly than by genetic adaptation (although
137 these behaviours can also be maladaptive; Barrett, Zepeda, Pollack, Munson, & Sih, 2019), with potential
138 to feedback on genetic change both within species and across communities if behavioural responses
139 become maintained in space and time (Whitehead, Laland, Rendell, Thorogood, & Whiten, 2019). Current
140 phenotypic responses may therefore reflect past phenotypic resistance or recovery, with or without
141 concomitant genetic change. As a consequence, harnessing knowledge about the past will likely require
142 integrating plasticity, epigenetics and genetic information (e.g. McNamara, Dall, Hammerstein, & Leimar,
143 2016) and has potential to provide a major advance across these fields.

144

145 At the level of the population or community, perturbations can reduce or increase individual or species
146 diversity, and thus impact any future response to disturbances. For example, past climatic fluctuations in
147 the Amazon basin have given rise to areas of more diverse avian fauna in the western parts compared to
148 the south-east. Thus, the south-eastern parts are expected to be more vulnerable to ongoing stress
149 posed by deforestation and climate change (Pontes-da-Silva et al., 2018). Community changes caused
150 by past disturbances may also determine subsequent community assembly through complex cascading
151 effects on species succession. For instance, when species re-colonize an area, or are reintroduced after
152 a perturbation, the order in which species arrive may be important for community assembly (i.e. priority
153 effect or founder control; Fukami, 2015) and future resilience. Disturbances may also fuel rapid evolution
154 of populations, which can, in turn, alter community assembly (Legrand et al., 2017). Discussions
155 regarding the role of genetic diversity and plasticity for resilience are therefore analogous to discussions
156 about how species diversity or functional diversity promotes resilience at the ecosystem-level, and also
157 how in some cases, turnover of species is necessary (e.g. Oliver et al. 2015). Determining whether the
158 effects of past diversity on present states are generalisable across biological levels will be an important
159 step in developing our understanding of biological resilience.

160

161

TEXT CONTINUED AFTER BOX 1

162

163

164 **Box 1: Integrating ecology and evolution across scales**

165 Evolutionary mechanisms (mutation, drift, migration, natural selection) generate changes in allele
166 frequencies from one generation to another (i.e. microevolution) and, given sufficient time or conditions,
167 can lead to large-scale changes that transcend species boundaries (i.e. macroevolution). Similarly,
168 processes that influence ecology (e.g. density, connectivity, competition, species interactions) at smaller
169 scales (e.g. within populations, communities) give rise to large-scale macroecological patterns (e.g.
170 biodiversity and ecosystem function). Darwin made no distinction between micro and macro scales, nor
171 did he separate ecology from evolutionary processes (see McGill et al., 2019). Over the 20th century,
172 however, research in ecology and evolution specialised to specific scales and processes (McGill et al.,
173 2019). Adopting a biological resilience framework, however, necessitates reintegration. How might this be
174 achieved?

175

176 Eco-evolutionary dynamics provides a potential solution to reintegrate ecological and evolutionary
177 processes across scales (Hendry, 2019; Ware et al., 2019; Bassar et al. 2021). Work in this rapidly
178 developing field is increasingly scaling up from population-level studies (Schoener, 2011) to analyse how
179 evolutionary processes impact ecological dynamics (and *vice versa*) in communities and even
180 ecosystems (Ware et al., 2019), with explicit acknowledgement that interactions and feedback can occur
181 across non-adjacent biological levels (e.g. see Figure 1 in Palkovacs & Hendry 2010 and in Ware et al.
182 2019). For example, a recent *in vitro* study propagated artificial bacterial communities of 34-species to
183 test how initial species-level traits and rapid genetic mutation influenced changes in community-level
184 species and genetic composition following pulses of antibiotic disturbance (Cairns, Jokela, Becks,
185 Mustonen, & Hiltunen, 2020). Although communities appeared to respond to the disturbances according
186 to classic processes of sorting by species' traits, rapid within-species evolution of antibiotic resistance
187 also occurred. Critically, these new variants persisted and left signatures of evolutionary change, despite
188 immigration of additional antibiotic-susceptible species and recovery of community composition. Studies
189 of eco-evolutionary dynamics are also beginning to expand in scope and take a landscape perspective
190 (Legrand et al., 2017; Nadeau & Urban, 2019; Tylianakis & Maia, 2020). Explicit consideration of habitat
191 fragmentation and climate change on both the ecological responsiveness and rapid evolution of dispersal

192 behaviour, for example, might resolve why some species are not experiencing range shifts as expected
193 (Nadeau & Urban, 2019). The role of evolutionary feedbacks on ecosystem-level processes is also now
194 beginning to attract attention, suggesting that evolutionary changes in the variation of traits may play an
195 important role in shaping how and when ecosystems reach tipping points and possibly irreversible
196 ecosystem change (Dakos et al., 2019).

197

198 At the macro scale, fusion of ecology and evolution has typically been limited (McGill et al., 2019). For
199 example, phylogenetic ecology attempts to integrate macroevolutionary patterns into studies of
200 community function, with studies suggesting that increased phylogenetic diversity can be critical for
201 ecosystem stability (Cadotte, Dinnage, & Tilman, 2012), but sometimes not (Winter, Devictor, &
202 Schweiger, 2013). There is much scope, however, for evolutionary history to provide further information
203 than as indicators of relatedness (Swenson, 2019). For example, Zitnik and colleagues compared protein
204 interactomes, complex networks of molecular interactions, across the tree of life to reveal how they evolve
205 greater resilience to a loss of network connections over time (Zitnik, Sosič, Feldman, & Leskovec, 2019).
206 At a different level of biological organisation, incorporating historical global temperature records, species-
207 level functional traits, and rates of phylogenetic diversification is also helping to explain how
208 microevolutionary history induces different macroevolutionary responses to temperature change across
209 angiosperms (Sun et al., 2020). Understanding biological resilience will require a step-change from
210 describing macro- or micro- scale patterns to demonstrating how evolutionary and ecological processes
211 shape short- and longer-term responses to environmental change. Bridging ecology and evolution across
212 these scales is still in its infancy (McGill et al., 2019) so adopting this framework could also provide
213 stimulus to return to a more Darwinian integrated approach.

214

215 END OF BOX 1

216

217 **Finding the right scale: Effects of perturbations vary across biological levels**

218 Although natural systems can also face novel perturbations to those experienced in the past (Donohue et
219 al., 2016), if we can uncover how elements of the system have responded to past perturbations, then this
220 information will become useful for predicting current and future change. However, perturbations
221 themselves can be complex and perturbations vary in intensity, duration, frequency and spatial extent,
222 and can, depending on their nature, cause gradual changes in ecosystem functions and services, or lead
223 to more drastic regime shifts (Barnosky et al., 2012). The impacts of perturbations can also vary across
224 biological levels; for example, an adaptive mutation enabling a species to exploit the perturbed ecosystem
225 can lead to community effects by outcompeting other species. Furthermore, perturbations are often
226 simultaneous and these may be related directly, such as warmer temperatures and increased droughts,
227 or indirectly, such as invasive species and eutrophication. The end result is often non-linear, with
228 simultaneous perturbations having synergistic effects (Brook, Sodhi, & Bradshaw, 2008) or generating
229 cascading processes (e.g. co-extinctions; Colwell, Dunn, & Harris, 2012). For example, communities often
230 cope with increasing disturbances with minimal apparent signs of stress, but then rapidly collapse when
231 the degree of perturbation reaches a tipping point (e.g. pollinator communities; Lever, van Nes, Scheffer,
232 & Bascompte, 2014). Although increasing theoretical and experimental work suggests that collapse in
233 natural systems can be anticipated by early warning signals (Scheffer et al., 2012), detecting these
234 signals in highly variable real world systems remains a great challenge (Dakos et al., 2019).

235

236 Adopting a biological resilience framework could help to predict these events as incorporating a longer
237 time horizon reveals resilience to be a dynamic and constantly evolving product of long term (co-)
238 evolutionary, ecological and biogeographical processes (e.g. Baruah, Clements & Ozgul, 2020).
239 Understanding how these processes operate at different biological levels of organisation is critical, as the
240 rate of evolution for example is constrained by generation times that vary from minutes (e.g. cells and
241 microbes) to centuries (e.g. trees), reproductive strategy influences opportunities for outcrossing and
242 mutation, and migration can diversify or limit local genotypic and phenotypic variation. However, at
243 present it remains unclear whether one level in particular will be of greater importance for predicting
244 responses to current and future perturbations. Furthermore, while it is likely that responses of one level to

245 a given perturbation will influence how multiple other levels respond, investigations into the carry-over
246 effects of perturbations across biological levels are few and mostly focus on adjacent levels (e.g. changes
247 in population influence response of communities, Strona et al. 2021). The composition, structure and
248 spatial context of a perturbed population or ecological community also needs to be taken into account
249 (Thebault & Fontaine, 2010). Range-edge populations, for example, can be comprised of a different set of
250 individual response-types than those found in the range core (e.g. spatial sorting; Massot, Legendre,
251 Fédérici, & Clobert, 2017) and potentially set up cascades of change across other biological levels (e.g.
252 reduced genetic diversity; Sgrò et al., 2011), while fragmented habitats influence the degree to which
253 species can reduce their exposure to perturbations by shifting, shrinking or expanding their range via
254 dispersal (Fahrig, 2003), or by modifying physiological or behavioural responses (Baguette & Van Dyck,
255 2007). Spatial context also has fundamental implications for longer-term adaptation to environmental
256 change as it shapes gene flow (Anderson et al., 2010). Integrating past and present distributions and
257 habitats is therefore likely to be a key, albeit challenging aspect. Nevertheless, using evolutionary history
258 as a 'natural experiment' and integrating information about adaptation explicitly into a resilience
259 framework could provide a previously untapped resource for predicting how ecological systems respond
260 to perturbations.

261

262 **Biological resilience framework generates testable hypotheses**

263 It is clear that determining how different biological levels resist and recover and buffer other levels from
264 perturbations will be complex, and that harnessing available information from the past is not
265 straightforward. However, theory and mathematical models lay the foundations for identifying what to
266 measure from experimental and empirical systems and how to extract these observations from real data.
267 For example, careful calibration of the effect of a perturbation with respect to the undisturbed state is
268 necessary to obtain common metrics that are comparable across biological levels and study systems
269 (Ingrisch & Bahn, 2018). Efforts to incorporate evolutionary perspectives into models of ecosystem-level
270 responses (e.g. tipping points, Dakos et al., 2019; warning signals, Baruah et al. 2020) are beginning, and
271 there is growing theory surrounding the ecological and evolutionary dynamics of resistance (e.g.
272 antibiotics, Meredith et al., 2018) and rapid genetic adaptation to ecological change (e.g. Waldvogel et al.,

273 2020). New theory is, however, required to bridge resistance and recovery responses across biological
274 levels. A long-term problem in ecological modelling is that theoretical models are good for understanding
275 causality, but difficult to test critically against data, whereas statistical models are correlative, and thus
276 may not identify the relevant underlying mechanisms even if they fit the present data well. Furthermore,
277 incorporating complex processes across many levels of biological organisation within one model is both
278 computationally and mathematically challenging.

279

280 Overcoming these shortcomings is especially critical for studying biological resilience, because it is likely
281 to often be driven by the interaction of many complex processes at many levels of biological organization.
282 Nevertheless, considering the effects of perturbations across biological levels in terms of eco-evolutionary
283 form and function helps generate hypotheses concerning the role of historical disturbances in shaping
284 future resilience: (i) past experience primes a biological entity to cope with future disturbances of a similar
285 nature. Alternatively, but not necessarily mutually exclusively, (ii) populations and communities exposed
286 to more variable environments and higher levels of disturbance over the long term are expected to be
287 most resilient. However, even these may accrue a resilience debt if the magnitude and frequency of the
288 disturbances differ too much from their historical disturbance regimes (Waples et al., 2009). Finally, (iii)
289 even without long-term disturbance histories, rapid adaptation may improve resilience against specific
290 stressors. This may, however, come at the cost of decreased resilience in the longer term because of
291 reduced pre-existing diversity after rapid adaptation or altered species interactions (Sgrò et al., 2011;
292 Stange, Barrett, & Hendry, 2020). Aspects of these hypotheses have already begun to be tested (Table
293 1), but not yet across biological levels within a relevant system.

294

295

296

297

TEXT CONTINUED AFTER TABLE 1

298 **TABLE 1.** Three hypotheses regarding how the ecological and evolutionary past shapes current and future
 299 responses to environmental change, and the multiple study approaches required to understand this biological
 300 resilience (with examples).

Hypotheses	Methodological approaches	Examples
(i) past experience primes a biological entity to cope best with future disturbances of a similar nature		
	Describe patterns using correlational or before-after survey data	<ul style="list-style-type: none"> • Current and future responses are mediated by past infection using long-term data on Soay sheep (Leivesley et al., 2019) • Co-occurrence of taxa before and after Holocene (Lyons et al., 2016)
	Use modelling and simulations to generate testable predictions	<ul style="list-style-type: none"> • Transgenerational priming (Kokko et al., 2017)
	Perform experimental perturbations in - cosms or field settings	<ul style="list-style-type: none"> • Experimental evolution with yeast (López-Maury et al., 2008) • Legacy effects of drought exposure on microbial communities (Krause et al., 2018) • Transgenerational acquired resistance in model plants (Holeski et al., 2012) • Resurrection studies (Franks et al., 2008)
	Interrogate findings with data from natural experiments	<ul style="list-style-type: none"> • Captive and wild songbirds respond differently to temperature perturbations (Verhagen et al., 2020)
(ii) diversity of environments and disturbances in the past generates greater resilience in the future		
	Make use of long-term survey data and/or big ecological and genetic	<ul style="list-style-type: none"> • Paleological history (Barnosky et al., 2017) • Ecological and evolutionary memory (Waples et al., 2009; Zitnik et al., 2019)

datasets (including ancient DNA) to measure past diversity	<ul style="list-style-type: none"> Adaptive genetic diversity (Sgrò et al., 2011)
Use modelling and simulations to generate testable predictions	<ul style="list-style-type: none"> Predicting a species response to environmental change when preadaptation of community differs (de Mazancourt, Johnson, & Barraclough, 2008)
Perform experimental perturbations in - cosms or field settings	<ul style="list-style-type: none"> Resurrection studies (Franks et al., 2008)
Interrogate findings with real-world examples, e.g. natural experiments	<ul style="list-style-type: none"> Biological invasions (Simberloff et al., 2013)

(iii) rapid adaptation to match current conditions reduces future resilience

Compare current resilience of biological entities and search for signs of rapid adaptation in the past	<ul style="list-style-type: none"> Genome-wide scans in forest trees to detect adaptation to aridity (Steane et al., 2014)
Use modelling and simulations to generate testable predictions	<ul style="list-style-type: none"> Evolutionary rescue (Bell, 2013)
Experimentally induce a novel perturbation in cases where rapid adaptation is present vs. absent	<ul style="list-style-type: none"> Resurrection studies (Franks et al., 2008)

302 **Approaches to understand biological resilience**

303 Understanding biological resilience will require concerted multidisciplinary research programmes where
304 the effects of a perturbation (or multiple stressors) in terms of resistance and recovery responses are
305 investigated across different levels, and where feedback among levels is also measured explicitly (Table
306 1, Figure 1). At present, research into coral reef resilience provides a worked example: surveys and
307 experiments have demonstrated that different coral species exhibit different degrees of resistance and
308 recovery to similar stressors (Hughes et al. 2010). Comparing the species' evolutionary history provides
309 some insight into why: a recent study suggests Caribbean corals show lower recovery than Indo-Pacific
310 corals due to an evolutionary bottleneck 2.8 million years ago that favoured large and long-lived species
311 with low rates of recruitment (Roff 2021). Efforts to investigate genomic predictors of coral bleaching
312 (Fuller et al. 2020), and even to assist evolution towards more resilient forms (van Oppen et al. 2015), are
313 also now attracting wide attention (Bay & Guerrero 2020). Furthermore, mapping dependencies of coral-
314 fish species based on natural history and fitting structural equation models has recently suggested that
315 coral loss may lead to substantial negative change in fish diversity and biomass worldwide, with effects
316 extending beyond the fish species directly dependent on corals (Strona et al. 2021).

317

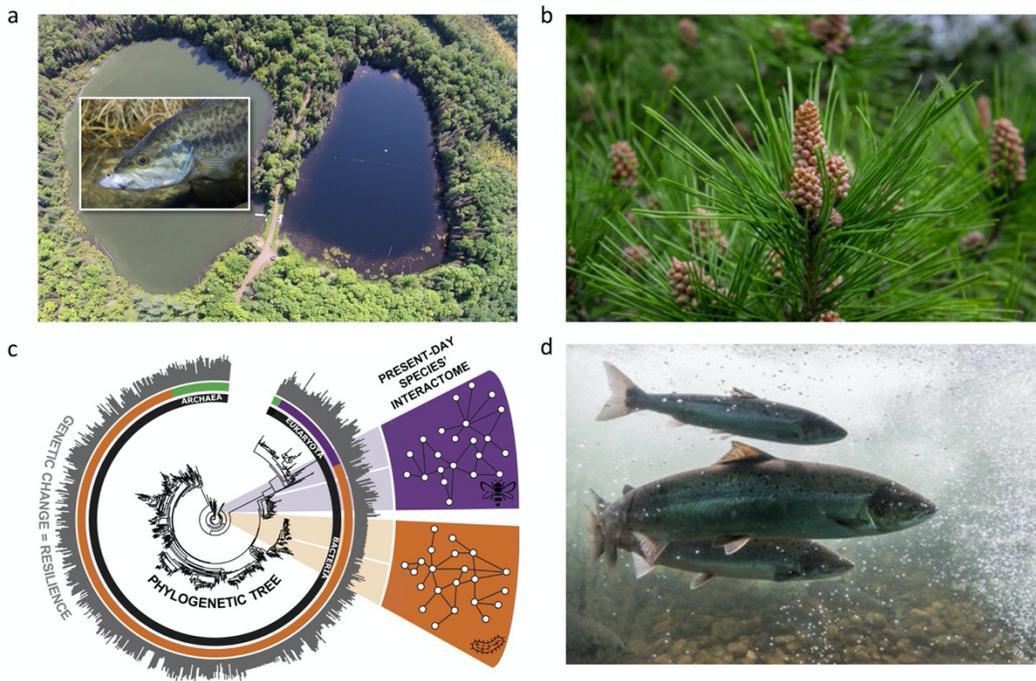
318 There are many other studies beyond this example that report genetic-, phenotypic-, or community-level
319 changes along environmental gradients or responses to natural changes, but far fewer either consider
320 more complex environmental scenarios (e.g. multiple or sequential stressors) or how the effects at one
321 biological level may affect others. As such, much of the current work in understanding biological resilience
322 (even if not yet couched in this terminology) relies on surveys and correlations that are carried out at one
323 level. For example, 'which genes contribute to more resilient phenotypes?' (Papakostas et al., 2012),
324 'which populations are more resilient to certain perturbations?' (Thom et al., 2019) or, 'which species are
325 most affected by which particular aspects of a perturbation?' (Strayer, 2010). Furthermore, the results of
326 experiments, particularly into resilience at the cellular (López-Maury, Marguerat, & Bähler, 2008) or
327 genetic levels (Kokko et al., 2017), are often not interpreted in a broader ecological context or compared
328 to available data from natural populations (Verhagen, Tomotani, Gienapp, & Visser, 2020). Here we
329 explore how we can move beyond studying the effects of single perturbations or single species or levels

330 and progress towards more complex experimental designs and assessments of more complex situations
331 in the wild.

332

333

TEXT CONTINUED AFTER FIGURE 2



334
 335 **FIGURE 2.** Examples of approaches to study biological resilience within biological levels (levels and
 336 approaches in bold): (a) **Communities & ecosystems:** Predators (inset) were introduced into a lake (left)
 337 in a **semi-natural experiment** and their abiotic and biotic effects were tracked over 3 years to test
 338 warning signals of a regime shift in an aquatic food web (compared to an undisturbed lake, right), as
 339 predicted by **modelling** after **long-term monitoring** (Carpenter et al., 2011). (b) **Individuals &**
 340 **populations:** Various **modelling** methods, coupled with data from **semi-natural experiment** genetic
 341 provenance trials for *Pinus sylvestris*, investigated how variation in population-level responses to
 342 environmental change (i.e. phenotypic plasticity and local adaptation) can influence species-level range
 343 expansion under climate change (Valladares et al., 2014). (c) **Cells & proteins:** Over 8 million protein-
 344 protein interactions from 1,840 species were **data mined** to **model** protein interactomes (examples are
 345 shown for a eukaryotic, purple, and bacterial, orange, species). Species' level evolutionary histories and
 346 ecological characteristics were then used to investigate how resilience varies at the protein level (Zitnik et
 347 al., 2019). (d) **Genes & genomes:** A **wild survey** of gene-linked loci and gene ontology information in
 348 *Salmo salar* populations tested the hypothesis that stronger signals of selection occur in loci with
 349 immune-related functions (Tonteri, Vasemägi, Lumme, & Primmer, 2010).

350 Image Credits: Fig.2a adapted to include inset image with permission from S.R. Carpenter, published
351 under the Creative Commons Attribution 4.0 International Public License (CC BY
352 4.0; <https://creativecommons.org/licenses/by/4.0/>). Fig. 2a(inset), Fig. 2b, Fig. 2c (eukaryote and
353 bacteria symbols), and Fig. 2d reproduced with permission from Adobe Stock (a(inset): rostislav –
354 stock.adobe.com, b: Quatrox Production - stock.adobe.com, c(eukaryote): vit003 - stock.adobe.com,
355 c(bacteria): vector_best - stock.adobe.com, d: Jakub Rutkiewicz – stock.adobe.com). These are not
356 covered by the CC-BY 4.0 licence and further reproduction requires permission from the copyright
357 holder. Fig. 2c is modified (rotated, labels added) with permission of the author from Figure 1E from
358 Zitnik et al. 2019; Proceedings of the National Academy of Sciences; Evolution of resilience in
359 protein interactomes across the tree of life.; <https://doi.org/10.1073/pnas.1818013116>; Published
360 and distributed under the terms of the Creative Commons Attribution-Non Commercial-No
361 Derivatives 4.0 License (CC-BY-ND 4/0); <https://creativecommons.org/licenses/by-nc-nd/4.0/>. It is
362 not covered by the CC-BY 4.0 licence.
363
364

365 To enable future studies to cover multiple biological levels, incorporating standardized collection of data
366 and sample material from multiple biological levels (e.g. genetic material, phenotype and community
367 structure) into geographical surveys and long-term studies is a good starting point. If these standardised
368 surveys are conducted over multiple seasons, years, or generations, this long-term monitoring has the
369 potential to facilitate (i) detection of subtle responses and/or subtle perturbations, (ii) replication over time,
370 and (iii) detection of ecological and evolutionary memories (Grant et al., 2017). The same
371 recommendation is relevant for “opportunistic” sampling following the (often unexpected) formation of a
372 resilience-relevant gradient/difference. Data for multiple biological levels at sites that have experienced a
373 heat wave for example, or an oil spill or chemical release, can either be compared to those of a nearby
374 site that did not experience the perturbation (Ellegren, Lindgren, Primmer, & Møller, 1997), or in the event
375 that surveys of the affected sites were conducted prior to the perturbation, a ‘before vs. after’ analysis can
376 be conducted (Bergen et al., 2020). Second, the prehistoric and paleoecological record is an important
377 potential source of survey data, as it is now becoming tractable to incorporate it with extant data (Fraser
378 et al., 2020). This paleo-perspective could offer natural experiments: data are available to potentially help
379 explain how community assembly (and disassembly) works when time spans are increased (Lyons et al.,
380 2016), for example, or how genetic structure and adaptations respond to perturbations ranging from major
381 extinctions to rapid climate change or species invasions over long time periods (e.g. Frisch et al. 2014).

382

383 A major challenge for survey approaches mentioned above however is to disentangle the effects of co-
384 varying environmental characteristics (e.g. photoperiod and temperature along a latitudinal gradient, or
385 simultaneous drought and reduced food availability). Therefore, experiments in semi-natural (e.g. *in vitro*
386 microcosms or outdoor mesocosm setups) or field settings (e.g. Figure 2a,b; ponds/tanks, forest/field
387 plots, enclosures suitable for small mammals, or free-ranging individuals and populations) are an
388 essential third approach to test how resilience occurs across biological levels, and offer an attractive
389 compromise where ‘real-world’ conditions are partly retained but where some manipulation and/or control
390 is nevertheless possible, together with replicates (Hendry, 2019). These experiments can range greatly
391 across organismal scale, geography, and biological levels (e.g. Figure 2b; Valladares et al., 2014), and
392 can also be conducted alongside interventions to mitigate species decline or change in ecosystem

393 function (e.g. conservation actions including introductions of individuals or translocations of populations,
394 Franks et al., 2020), if the selection of individuals or species to be moved is designed to test the relative
395 resilience of different characteristics (e.g. social behaviour: Goldenberg et al., 2019, genetic diversity:
396 Stange et al., 2020). Although further removed from 'real world' conditions, common garden experiments
397 (i.e. the rearing individuals in a controlled environment under common conditions) could be used to study
398 responses to environmental or anthropogenic stressors by adding 'treatments' such as thermal stress,
399 disease, or changes in community (e.g. flour beetles: Koch & Guillaume, 2020; burying beetles: Sun &
400 Kilner 2020). Here, environmental differences can be eliminated or specific environmental factors can be
401 tested so that the extent of resilience that is plastic versus evolutionary (e.g. fish: Papakostas et al., 2014,
402 crops: Bustos-Korts et al., 2019) can be measured. Resurrection-type experiments are also a promising
403 approach in some taxa, when genotypes that have experienced varying conditions are available for tests
404 under experimental conditions (Franks et al., 2008). Experimental designs like these outlined above have
405 been criticised for over-simplifying ecological processes, however taking an experimental approach will
406 be essential to tease apart the relative effects of multiple stressors, either simultaneously, or sequentially,
407 or at different stages of an organism's life-history. Results of simpler experimental designs or studies at
408 single biological levels may also enable refining hypotheses and study designs for the future study of
409 other biological levels in more complex conditions.

410

411 Fourth, eco-evolutionary and environmental Big Data, from the molecular to the ecosystem level, provides
412 a broad and expanding scope, particularly when datasets span space and/or time. At the molecular level,
413 Big Data on genes and genomes (NCBI; National Library of Medicine (US) & National Center for
414 Biotechnology Information, 1988) and their function (Gene Ontology (GO; Ashburner et al., 2000) and
415 Kyoto Encyclopaedia of Genes and Genomes (KEGG; Kanehisa, Sato, Furumichi, Morishima, & Tanabe,
416 2019) databases are rapidly increasing. These databases are designed to be taxonomically comparable,
417 or even species-neutral, to enable transfer of functional annotation (molecular function, biological role and
418 cellular location) or gene network information derived from model organisms to inferred orthologues in
419 newly sequenced species. If the current focus on medical science or morphological characters broadens
420 to encompass functions in response to ecological stimuli (Primmer et al., 2013), then big genomic data

421 will become an even more useful resource for studying the molecular basis of biological resilience (e.g.
422 Figure 2c,d). Similarly, finding the most potent data sources for reconstructing time series into the past
423 still requires innovation, but this approach carries considerable promise for analyses of resilience to
424 changes that have already occurred. For example, abiotic data from the last few decades are now openly
425 available (e.g. CORINE (Copernicus Land Monitoring Service, 2018), USGS (USGS, 2020), WorldClim
426 (Fick & Hijmans, 2017)) and big data on species occurrences (GBIF (GBIF.org, 2020)), traits (TRY
427 (Kattge et al., 2020), Coral Trait Database (Madin et al., 2016)) and abundances through time
428 (Ovaskainen et al., 2020) are becoming available at an increasing rate. Collecting data of changes in the
429 deeper past requires continued efforts in digitising physical collections (museum specimens; The NOW
430 Community, 2020) and application and development of new techniques for data extraction and analysis
431 (Fraser et al., 2020).

432

433 As the resolution and density of data increases, and new algorithms that make use of large-scale
434 computational resources become available, the possibilities to find and match comparable drivers-to-
435 biotic-units cases will increase. However, most of the global databases at present contain (partially) non-
436 comparable data, and experimental data are rarely combined with observational data despite potential to
437 increase credibility of conclusions (Kotta et al., 2019). Existing data can be analysed by taking advantage
438 of newly developed methods that minimise biases in unrelated or uncertain data (e.g. Bayesian
439 approaches; Ovaskainen & Abrego, 2020), or when fully comparable data are available, by using
440 mechanistic models that allow moving beyond correlative analyses (e.g. individual-based models;
441 DeAngelis & Grimm, 2014). Any data analysis must, however, be based on theoretically sound models as
442 blindly applying black-box machine learning algorithms to interpret data may lead to conclusions that are
443 not biologically sensible (Hartmann, et al., 2017). Moreover unbalanced sampling may lead to incorrect
444 interpretations if not accounted for in analyses (Foster et al., 2021) – a problem similar to discriminatory
445 biases in social data applications of machine learning. Other areas of artificial intelligence, such as
446 symbolic regression (Udrescu & Tegmark, 2020), hold much promise to improve our ability to predict the
447 consequences of ongoing and future change as they can provide both power and interpretability of
448 natural laws.

449

450 **Translating biological resilience from research to management and conservation**

451 While there have been many calls to adapt management and conservation of natural resources to
452 improve resilience to environmental change, substantial obstacles remain before this can be realised.
453 First, managers require indicators at levels most appropriate for decision-making. Many of the indicators
454 currently available, however, are system-wide or remain challenging to quantify (Dunbar et al., 2020;
455 Ingrisch & Bahn, 2018; Standish et al., 2014). Indicators based on species diversity and habitat
456 connectivity, for example, allow assessment of large-scale patterns (Dunbar et al., 2020), but they are
457 less helpful for management of more tractable system components. Similarly, current discussions around
458 genetic diversity are often difficult to reconcile with ecosystem health as they operate at different
459 timescales (but see Kettenring et al. 2014 for a notable example in plants). Second, attempts to manage
460 'for resilience' typically focus on avoiding thresholds or tipping points. Rather, managers need to compare
461 alternative choices, assess potential outcomes with greater certainty than is currently possible, and
462 manage adaptively (Weise et al., 2020). Third, management approaches largely aim for current or
463 historical states, rather than attempting to forecast outcomes according to novel future conditions. This is
464 especially problematic when the time horizon is long (Weise et al., 2020), for example in forestry and
465 agriculture where long or uncertain time horizons play a large part in the difficulty to translate
466 recommendations (Dhankher & Foyer, 2018; Millar, Stephenson, & Stephens, 2007). Determining how
467 resilience operates at different biological levels has potential to move beyond this stalemate, by using the
468 ecological and evolutionary history of components of the system (Fraser et al., 2020; Yang et al., 2019) to
469 better predict future states under different management scenarios (e.g. Box 2).

470

471

TEXT CONTINUED AFTER BOX 2

472

473

474 **BOX 2.**

475 Here we highlight the broad potential for the applicability of a biological resilience approach by briefly
476 exploring how it could influence translation and management in two divergent examples: (i) forestry and
477 agriculture, and (ii) human health.

478

479 **(i) Biological resilience in forestry and agriculture**

480 In the past, forest managers have assumed that the climate and other associated factors will remain
481 stable, in spite of the long generation times and individual lifespans of many forest tree species and
482 biomes (Millar et al., 2007). However, soil degradation (for example) can occur rapidly compared to the
483 lifespan of the forest and then impact on the ability of trees to withstand other environmental perturbations
484 (Swinfield et al., 2020). Similarly, modern plant breeding selects for yield potential under high and stable
485 resource supply, and generally relies on genetically uniform cultivars. A biological resilience framework,
486 however, encourages a different approach. For example, studies of local adaptation at the population
487 level would help to understand how we can best buffer food and/or timber production against
488 perturbations, perhaps by combining long-term data series and targeted experiments informed by
489 historical farming practice or evolutionary processes (Millar et al., 2007). In a context with clear
490 applications for management, Ives and colleagues recently discovered that spatial heterogeneity in crop-
491 harvesting is a major driver of the ecological and evolutionary feedbacks that limit resistance of pea
492 aphids to parasitoid wasps, an important biological control agent (Ives et al., 2020). Past perturbations
493 also leave abiotic 'stress memory', encoded in DNA methylation and chromatin marks, which may
494 increase resilience over multiple generations (Chang et al., 2020; Friedrich, Faivre, Bäurle, & Schubert,
495 2019) in a process of acquired transgenerational resistance (Holeski et al., 2012). Similarly, interactions
496 across trophic and biological levels are well-known features of plant growth and health, with key work
497 demonstrating that these also influence resilience (e.g. plant-microbe interactions influence resistance to
498 climate change; Rudgers et al., 2020). Harnessing this information could lead to improved crop plant and
499 tree breeding programmes (e.g. Messier et al., 2019), but much of this work remains embedded in model
500 plant systems, such as *Arabidopsis*. Understanding which features at what biological level are most

501 important to manage (e.g. managing for genetic diversity of monotypic plantations versus diversity of
502 associated mycorrhizal fungi) will require combined approaches and translation of work from model
503 species to natural systems.

504

505 **(ii) Biological resilience in human health**

506 While ecological systems are increasingly becoming viewed as socio-ecological systems (Ellis, 2015), the
507 idea that the human mind and body can be viewed as a complex ecological system is only just beginning
508 to be recognised (Bernstein, 2019). Understanding how circadian misalignment of sleep/wake cycles
509 leads to a mismatch between abiotic cues and internal cellular functions (e.g. impairment of beta cell
510 function and insulin sensitivity; Mason, Qian, Adler, & Scheer, 2020), and then scales up to affect system
511 health via resilience to disease and other stressors, could help to provide more appropriate guidelines for
512 managing shift work, for example. Recent experiences with COVID-19 also demonstrate the need to
513 consider how resilience operates across biological levels: identifying what makes an individual more
514 resilient to a virus at the cellular level (e.g. vaccine development) is not enough if insufficient people take
515 up the vaccine (i.e. population level), or if the virus itself evolves resistance. Indeed, understanding the
516 biological resilience of viral infections, or cancerous growths for example, to medical interventions could
517 assist in progress with treatment. Genetic heterogeneity is known to negatively affect treatment success
518 in cancer (Maley et al., 2006), yet this heterogeneity reflects the selective pressures endured, and the
519 variation accumulated, during the whole history of that cancer and can reveal vulnerabilities to therapy
520 (Alexandrov, Nik-Zainal, Siu, Leung, & Stratton, 2015). Furthermore, life-history strategies of cells, such
521 as dormancy, can blunt the effects of therapy (e.g. tuberculosis). This suggests that diversity could be an
522 important component of resilience in human health, but this requires testing in translational models.

523

524

525 END OF BOX 2

526

527

528 **Challenges of implementing a biological resilience framework**

529 Here we have argued that understanding and managing for biological resilience requires moving away
530 from the approach of considering function or resilience only at the level of ecosystems, or of focusing
531 studies within a single biological level. We have also stressed how the resilience of the present state not
532 only relies on perturbations experienced in the past (whether contemporary, transgenerational, or deeper
533 in evolutionary time) but that we can also access information about these past responses. Nevertheless,
534 incorporating evolutionary history and complex interactions within and across biological levels is non-
535 trivial, and key challenges exist for modelling complexity and broadening the scope of data collection, as
536 well as setting the temporal and spatial boundaries of the systems or components being studied.

537

538 Firstly, in both theoretical and empirical work, we need to identify which connections among what levels
539 are most critical to study. A top-down view of ecosystems works best when considering change over a
540 relatively short period of time, and reduces power for forecasting future responses, either to predicted
541 environmental change or potential management interventions. In ecosystem ecology, species, for
542 example, are normally classified into functional types that leave out valuable information about
543 evolutionary responses to specific perturbations in the past. These responses can however be searched
544 for by mining existing data (e.g. Figure 2c; Zitnik et al., 2019) or by experiment (e.g. Oostra et al., 2018).
545 Similarly, we need to move beyond research focusing on what makes an individual, or a species,
546 resistant or tolerant to some perturbation without assessing its relevance to systems or communities.
547 Research in eco-evolutionary dynamics is already beginning to tackle these interactions (Box 1), and
548 adapting this approach to investigate resilience provides a model for moving forwards. While it is not
549 tractable to measure everything, well-controlled experiments can provide critical data to understand the
550 mechanisms that drive biological resilience – or the lack of it. However, as experiments entail at least
551 some simplification of natural complexity, results will need to be linked conceptually to surveys of the
552 relevant organisms and ecosystems.

553

554 Considering multiple levels of biological organisation will also necessitate data collection that tracks
555 responses and maximises phylogenetic, functional, spatial and temporal coverage with minimum

556 monetary cost (Cardoso & Leather, 2019). This is a challenging task for independent research groups as
557 the acquisition of uninterrupted and consistent time series of ecological and environmental data depend
558 on continued funding. Therefore, coordinated multidisciplinary research projects would enhance data
559 collection and optimise funding streams, making it possible to expand the scope from single- to multiple
560 levels. Some types of data are already available to inform about responses to past conditions, but if we
561 are to make better use of existing and future available datasets, these will require high quality metadata
562 annotations including as many potential ecological variables as possible (and not only the ones directly
563 related to the analyses data were collected for) and easy and open access (e.g. following the FAIR
564 principles; Wilkinson et al., 2016).

565

566 Providing the evidence necessary to make the case to policy makers is perhaps the most important
567 challenge. For example, accumulating knowledge on ecosystem resilience is yet to change the principles
568 of forestry or cropland management dramatically, which is alarming given that we know many current
569 management practices compromise the ability of future generations to meet their own needs. This may be
570 because resilience is currently difficult to quantify, and a lack of resilience is easier to recognise than a
571 successful management practice. A biological resilience framework could improve identification of
572 'resilience indicators' at scales in which management decisions are made. Tracking genetic diversity at a
573 species level, for example, is a feasible method to collect robust data, and could enable modelling of
574 which actions are likely to be most successful. A critical further step, however, will be improved monitoring
575 of the impact of potential indicators so that we are able to learn from both successful and less successful
576 implementations. Similarly, there are still substantial gaps to bridge between scientists, policymakers and
577 other stakeholders. For example, in commercial farming and forestry widespread adoption of science-led
578 practices depends on short-term economic benefits, so adoption will require policy-based incentives. A
579 deeper understanding of management practices, and co-creation of research questions with stakeholders
580 that will apply management practices, is essential, particularly if we are to implement decisions using an
581 experimental approach.

582

583 In summary, biological resilience requires shifting our perspective in eco-evolutionary studies towards
584 investigating terms of resistance versus recovery (the key conceptual outcomes in ecosystem resilience)
585 while also incorporating an eco-evolutionary perspective to better understand ecosystem-level processes
586 (Figure 1, Box 1). This requires real multidisciplinary coordinated actions. But we can also begin to take
587 small steps within existing research programmes. Researchers should consider reframing current
588 research to test theory regarding types of responses to perturbations under study. Or, we could consider
589 how influences from evolutionary history may impact ecological responses being detected under current
590 conditions. Although challenging, this approach should provide the advances in data collection, modelling,
591 and testing of hypotheses across levels that are urgently needed to improve resilience in the face of
592 current and future environmental challenges.

593

594 **Author contributions**

595 All authors were involved in Conceptualisation, Writing – original draft preparation, and Writing – review &
596 editing. R.T. and V.M. were responsible for preparation of the final version.

597

598 **Acknowledgements**

599 This manuscript is a contribution by members of the HiLIFE (Helsinki Institute for Life Science) Grand
600 Challenge programme in Understanding Biological Resilience (BIORESILIENCE), established after
601 external review by the HiLIFE Scientific Council and funded by the Academy of Finland funding
602 instrument PROF11 (awarded to the University of Helsinki). We are grateful to Jenni Villa and Unni
603 Pulliainen for their project coordination efforts.

604

605 **REFERENCES**

- 606 1. Alexandrov, L. B., Nik-Zainal, S., Siu, H. C., Leung, S. Y., & Stratton, M. R. (2015). A mutational
607 signature in gastric cancer suggests therapeutic strategies. *Nature Communications*, 6 (1), 8683.
608 <https://doi.org/10.1038/ncomms9683>
- 609 2. Anderson, C. D., Epperson, B. K., Fortin, M.-J., Holderegger, R., James, P. M. A., Rosenberg, M. S.,
610 ... Spear, S. (2010). Considering spatial and temporal scale in landscape-genetic studies of gene
611 flow. *Molecular Ecology*, 19 (17), 3565–3575. <https://doi.org/10.1111/j.1365-294X.2010.04757.x>
- 612 3. Ashburner, M., Ball, C. A., Blake, J. A., Botstein, D., Butler, H., Cherry, J. M., ... Sherlock, G. (2000).
613 Gene Ontology: tool for the unification of biology. *Nature Genetics*, 25 (1), 25–29.
614 <https://doi.org/10.1038/75556>
- 615 4. Baguette, M., & Van Dyck, H. (2007). Landscape connectivity and animal behavior: functional grain
616 as a key determinant for dispersal. *Landscape Ecology*, 22 (8), 1117–1129.
617 <https://doi.org/10.1007/s10980-007-9108-4>
- 618 5. Barnosky, A. D., Hadly, E. A., Bascompte, J., Berlow, E. L., Brown, J. H., Fortelius, M., ... Smith, A.
619 B. (2012). Approaching a state shift in Earth's biosphere. *Nature*, 486 (7401), 52–58.
620 <https://doi.org/10.1038/nature11018>
- 621 6. Barnosky, A. D., Hadly, E. A., Gonzalez, P., Head, J., Polly, P. D., Lawing, A. M., ... Zhang, Z.
622 (2017). Merging paleobiology with conservation biology to guide the future of terrestrial ecosystems.
623 *Science*, 355 (6325), eaah4787. <https://doi.org/10.1126/science.aah4787>
- 624 7. Barrett, B., Zepeda, E., Pollack, L., Munson, A., & Sih, A. (2019). Counter-Culture: Does Social
625 Learning Help or Hinder Adaptive Response to Human-Induced Rapid Environmental Change?
626 *Frontiers in Ecology and Evolution*, 7 (May), 1–18. <https://doi.org/10.3389/fevo.2019.00183>
- 627 8. Bartholomé, J., Brachi, B., Marçais, B., Mougou-Hamdane, A., Bodénès, C., Plomion, C., ... Desprez-
628 Loustau, M. (2020). The genetics of exapted resistance to two exotic pathogens in pedunculate oak.
629 *New Phytologist*, 226 (4), 1088–1103. <https://doi.org/10.1111/nph.16319>
- 630 9. Baruah, G., Clements, C. F., & Ozgul, A. (2020). Eco-evolutionary processes underlying early
631 warning signals of population declines. *Journal of Animal Ecology*, 89(2), 436–448.
632 <https://doi.org/10.1111/1365-2656.13097>

- 633 10. Bassar, R. D., Coulson, T., Travis, J. & Reznick, D. N. (2021). Towards a more precise – and
634 accurate – view of eco-evolution. *Ecology Letters*, 24 (4), 623–625.
- 635 11. Bay, R. A., Guerrero, L. (2020). Can genomes predict coral bleaching? *Science*, 369 (6501), 249–50.
636 <https://doi/10.1126/science.abc9342>
- 637 12. Bell, G. (2013). Evolutionary rescue and the limits of adaptation. *Philosophical Transactions of the*
638 *Royal Society B: Biological Sciences*, 368 (1610), 20120080. <https://doi.org/10.1098/rstb.2012.0080>
- 639 13. Bergen, E., Dallas, T., DiLeo, M. F., Kahilainen, A., Mattila, A. L. K., Luoto, M., & Saastamoinen, M.
640 (2020). The effect of summer drought on the predictability of local extinctions in a butterfly
641 metapopulation. *Conservation Biology*, 34 (6), 1503–1511. <https://doi.org/10.1111/cobi.13515>
- 642 14. Bernstein, A. (2019). All creatures great and small. *BMJ*, l2385. <https://doi.org/10.1136/bmj.l2385>
- 643 15. Bethesda (MD): National Library of Medicine (US), & National Center for Biotechnology Information.
644 (1988). National Center for Biotechnology Information (NCBI)[Internet]. Retrieved July 6, 2020, from
645 <https://www.ncbi.nlm.nih.gov/>
- 646 16. Brook, B., Sodhi, N., & Bradshaw, C. (2008). Synergies among extinction drivers under global
647 change. *Trends in Ecology & Evolution*, 23 (8), 453–460. <https://doi.org/10.1016/j.tree.2008.03.011>
- 648 17. Bustos-Korts, D., Dawson, I. K., Russell, J., Tondelli, A., Guerra, D., Ferrandi, C., ... van Eeuwijk, F.
649 A. (2019). Exome sequences and multi-environment field trials elucidate the genetic basis of
650 adaptation in barley. *The Plant Journal*, 99 (6), 1172–1191. <https://doi.org/10.1111/tpj.14414>
- 651 18. Cadotte, M. W., Dinnage, R., & Tilman, D. (2012). Phylogenetic diversity promotes ecosystem
652 stability. *Ecology*, 93 (sp8), S223–S233. <https://doi.org/10.1890/11-0426.1>
- 653 19. Cairns, J., Jokela, R., Becks, L., Mustonen, V., & Hiltunen, T. (2020). Repeatable ecological
654 dynamics govern the response of experimental communities to antibiotic pulse perturbation. *Nature*
655 *Ecology and Evolution*, 4 (10), 1385–1394. <https://doi.org/10.1038/s41559-020-1272-9>
- 656 20. Capdevila, P., Stott, I., Oliveras Menor, I., Stouffer, D. B., Raimundo, R. L. G., White, H., Barbour, M.,
657 & Salguero-Gómez, R. (2021). Reconciling resilience across ecological systems, species and
658 subdisciplines. *Journal of Ecology*, 109(9), 3102–3113. <https://doi.org/10.1111/1365-2745.13775>
- 659 21. Cardoso, P., & Leather, S. R. (2019). Predicting a global insect apocalypse. *Insect Conservation and*
660 *Diversity*, 12 (4), 263–267. <https://doi.org/10.1111/icad.12367>

- 661 22. Carpenter, S. R., Cole, J. J., Pace, M. L., Batt, R., Brock, W. A., Cline, T., ... Weidel, B. (2011). Early
662 warnings of regime shifts: A whole-ecosystem experiment. *Science*, 332 (6033), 1079–1082.
663 <https://doi.org/10.1126/science.1203672>
- 664 23. Carroll, S. P., Jørgensen, P. S., Kinnison, M. T., Bergstrom, C. T., Denison, R. F., Gluckman, P., ...
665 Tabashnik, B. E. (2014). Applying evolutionary biology to address global challenges. *Science*, 346
666 (6207), 1245993–1245993. <https://doi.org/10.1126/science.1245993>
- 667 24. Chang, Y., Zhu, C., Jiang, J., Zhang, H., Zhu, J., & Duan, C. (2020). Epigenetic regulation in plant
668 abiotic stress responses. *Journal of Integrative Plant Biology*, 62 (5), 563–580.
669 <https://doi.org/10.1111/jipb.12901>
- 670 25. Colwell, R. K., Dunn, R. R., & Harris, N. C. (2012). Coextinction and persistence of dependent
671 species in a changing world. *Annual Review of Ecology, Evolution, and Systematics*, 43 (1), 183–203.
672 <https://doi.org/10.1146/annurev-ecolsys-110411-160304>
- 673 26. Copernicus Land Monitoring Service, E. E. A. (2018). CORINE land cover. Retrieved from
674 <https://land.copernicus.eu/pan-european/corine-land-cover>
- 675 27. Dakos, V., Matthews, B., Hendry, A. P., Levine, J., Loeuille, N., Norberg, J., ... De Meester, L. (2019).
676 Ecosystem tipping points in an evolving world. *Nature Ecology and Evolution*, 3 (3), 355–362.
677 <https://doi.org/10.1038/s41559-019-0797-2>
- 678 28. de Mazancourt, C., Johnson, E., & Barraclough, T. G. (2008). Biodiversity inhibits species'
679 evolutionary responses to changing environments. *Ecology Letters*, 11 (4), 380–388.
680 <https://doi.org/10.1111/j.1461-0248.2008.01152.x>
- 681 29. DeAngelis, D. L., & Grimm, V. (2014). Individual-based models in ecology after four decades.
682 *F1000Prime Reports*, 6, 39. <https://doi.org/10.12703/P6-39>
- 683 30. Desai, M. M. (2009). Reverse evolution and evolutionary memory. *Nature Genetics*, 41 (2), 142–143.
684 <https://doi.org/10.1038/ng0209-142>
- 685 31. Dhankher, O. P., & Foyer, C. H. (2018). Climate resilient crops for improving global food security and
686 safety. *Plant, Cell & Environment*, 41 (5), 877–884. <https://doi.org/10.1111/pce.13207>
- 687 32. Donelson, J. M., Sunday, J. M., Figueira, W. F., Gaitán-Espitia, J. D., Hobday, A. J., Johnson, C. R.,
688 Leis, J. M., Ling, S. D., Marshall, D., Pandolfi, J. M., Pecl, G., Rodgers, G. G., Booth, D. J., &

- 689 Munday, P. L. (2019). Understanding interactions between plasticity, adaptation and range shifts in
690 response to marine environmental change. *Philosophical Transactions of the Royal Society B:
691 Biological Sciences*, 374(1768), 20180186. <https://doi.org/10.1098/rstb.2018.0186>
- 692 33. Donohue, I., Hillebrand, H., Montoya, J. M., Petchey, O. L., Pimm, S. L., Fowler, M. S., ... Yang, Q.
693 (2016). Navigating the complexity of ecological stability. *Ecology Letters*, 19 (9), 1172–1185.
694 <https://doi.org/10.1111/ele.12648>
- 695 34. Dunbar, W., Subramanian, S. M., Matsumoto, I., Natori, Y., Dublin, D., Bergamini, N., ... Mock, G.
696 (2020). Lessons learned from application of the “Indicators of resilience in socio-ecological production
697 landscapes and seascapes (SEPLS)” under the Satoyama Initiative. In O. Saito, S. M. Subramanian,
698 S. Hashimoto, & K. Takeuchi (Eds.), *Managing socio-ecological production landscapes and
699 seascapes for sustainable communities in Asia: mapping and navigating stakeholders, policy and
700 action* (pp. 93–116). Singapore: Springer Singapore. https://doi.org/10.1007/978-981-15-1133-2_6
- 701 35. Ellegren, H., Lindgren, G., Primmer, C. R., & Møller, A. P. (1997). Fitness loss and germline
702 mutations in barn swallows breeding in Chernobyl. *Nature*, 389 (6651), 593–596.
703 <https://doi.org/10.1038/39303>
- 704 36. Ellis, E. C. (2015). Ecology in an anthropogenic biosphere. *Ecological Monographs*, 85 (3), 287–331.
705 <https://doi.org/10.1890/14-2274.1>
- 706 37. Fahrig, L. (2003). Effects of habitat fragmentation on biodiversity. *Annual Review of Ecology,
707 Evolution, and Systematics*, 34 (1), 487–515.
708 <https://doi.org/10.1146/annurev.ecolsys.34.011802.132419>
- 709 38. Fick, S. E., & Hijmans, R. J. (2017). WorldClim 2: new 1-km spatial resolution climate surfaces for
710 global land areas. *International Journal of Climatology*, 37 (12), 4302–4315.
711 <https://doi.org/10.1002/joc.5086>
- 712 39. Foley, J. A. (2005). Global consequences of land use. *Science*, 309 (5734), 570–574.
713 <https://doi.org/10.1126/science.1111772>
- 714 40. Foster, S. D., Vanhatalo, J., Trenkel, V. M., Schulz, T., Lawrence, E., Przeslawski, R., & Hosack, G.
715 R. (2021). Effects of ignoring survey design information for data reuse. *Ecological Applications*, 31
716 (6), e02360. <https://doi.org/10.1002/eap.2360>

- 717 41. Franks, S. J., Avise, J. C., Bradshaw, W. E., Conner, J. K., Etterson, J. R., Mazer, S. J., ... Weis, A.
718 E. (2008). The Resurrection Initiative: Storing Ancestral Genotypes to Capture Evolution in Action.
719 *BioScience*, 58 (9), 870–873. <https://doi.org/10.1641/B580913>
- 720 42. Franks, V. R., Andrews, C. E., Ewen, J. G., McCready, M., Parker, K. A., & Thorogood, R. (2020).
721 Changes in social groups across reintroductions and effects on post-release survival. *Animal*
722 *Conservation*, 23(4), 443–454. <https://doi.org/10.1111/acv.12557>
- 723 43. Fraser, D., Soul, L. C., Tóth, A. B., Balk, M. A., Eronen, J. T., Pineda-Munoz, S., ... Lyons, S. K.
724 (2020). Investigating biotic interactions in deep time. *Trends in Ecology & Evolution*,
725 <https://doi.org/10.1016/j.tree.2020.09.001>. <https://doi.org/10.1016/j.tree.2020.09.001>
- 726 44. Friedrich, T., Faivre, L., Bäurle, I., & Schubert, D. (2019). Chromatin-based mechanisms of
727 temperature memory in plants. *Plant, Cell & Environment*, 42 (3), 762–770.
728 <https://doi.org/10.1111/pce.13373>
- 729 45. Frisch, D., Morton, P. K., Chowdhury, P. R., Culver, B. W., Colbourne, J. K., Weider, L. J., &
730 Jeyasingh, P. D. (2014). A millennial-scale chronicle of evolutionary responses to cultural
731 eutrophication in *Daphnia*. *Ecology Letters*, 17(3), 360–368. <https://doi.org/10.1111/ele.12237>
- 732 46. Fukami, T. (2015). Historical contingency in community assembly: integrating niches, species pools,
733 and priority effects. *Annual Review of Ecology, Evolution, and Systematics*, 46 (1), 1–23.
734 <https://doi.org/10.1146/annurev-ecolsys-110411-160340>
- 735 47. Fuller, Z. L., Mocellin, V. J. L., Morris, L. A., Cantin, N., Shepherd, J., Sarre, L., et al. (2020).
736 Population genetics of the coral *Acropora millepora*: Toward genomic prediction of bleaching.
737 *Science*, 369 (6501), eaba4674. <https://doi.org/10.1126/science.aba4674>
- 738 48. GBIF.org. (2020). GBIF Home page. Retrieved June 11, 2020, from <https://www.gbif.org>
- 739 49. Gladstone-Gallagher, R. V., Pilditch, C. A., Stephenson, F., & Thrush, S. F. (2019). Linking traits
740 across ecological scales determines functional resilience. *Trends in Ecology and Evolution*, 34(12),
741 1080–1091. <https://doi.org/10.1016/j.tree.2019.07.010>
- 742 50. Goldenberg, S. Z., Owen, M. A., Brown, J. L., Wittemyer, G., Oo, Z. M., & Leimgruber, P. (2019).
743 Increasing conservation translocation success by building social functionality in released populations.
744 *Global Ecology and Conservation*, 18, e00604. <https://doi.org/10.1016/j.gecco.2019.e00604>

- 745 51. Grant, P. R., Grant, B. R., Huey, R. B., Johnson, M. T. J., Knoll, A. H., & Schmitt, J. (2017). Evolution
746 caused by extreme events. *Philosophical Transactions of the Royal Society B: Biological Sciences*,
747 372 (1723), 20160146. <https://doi.org/10.1098/rstb.2016.0146>
- 748 52. Hartmann, M., Hosack, G. R., Hillary, R. M., & Vanhatalo, J. (2017). Gaussian process framework for
749 temporal dependence and discrepancy functions in Ricker-type population growth models. *The*
750 *Annals of Applied Statistics*, 11 (3), 1375–1402. <https://doi.org/10.1214/17-AOAS1029>
- 751 53. Hendry, A. P. (2019). A critique for eco-evolutionary dynamics. *Functional Ecology*, 33 (1), 84–94.
752 <https://doi.org/10.1111/1365-2435.13244>
- 753 54. Hodgson, D., McDonald, J. L. & Hosken, D. J. (2015). What do you mean, 'resilient'? *Trends Ecology*
754 *Evolution*, 30 (9), 503–506.
- 755 55. Holeski, L. M., Jander, G., & Agrawal, A. A. (2012). Transgenerational defense induction and
756 epigenetic inheritance in plants. *Trends in Ecology & Evolution*, 27 (11), 618–626.
757 <https://doi.org/10.1016/j.tree.2012.07.011>
- 758 56. Holling, C. S. (1973). Resilience and stability of ecological systems. *Annual Review of Ecology and*
759 *Systematics*, 4 (1), 1–23. <https://doi.org/10.1146/annurev.es.04.110173.000245>
- 760 57. Hughes, T. P., Graham, N. A., Jackson, J. B., Mumby, P. J., & Steneck, R. S. (2010). Rising to the
761 challenge of sustaining coral reef resilience. *Trends in Ecology & Evolution*, 25 (11), 633–642.
- 762 58. Ingrisch, J., & Bahn, M. (2018). Towards a Comparable Quantification of Resilience. *Trends in*
763 *Ecology & Evolution*, 33 (4), 251–259. <https://doi.org/10.1016/j.tree.2018.01.013>
- 764 59. Ives, A. R., Barton, B. T., Penczykowski, R. M., Harmon, J. P., Kim, K. L., Oliver, K., & Radeloff, V. C.
765 (2020). Self-perpetuating ecological–evolutionary dynamics in an agricultural host–parasite system.
766 *Nature Ecology and Evolution*, 4 (5), 702–711. <https://doi.org/10.1038/s41559-020-1155-0>
- 767 60. Johnstone, J. F., Allen, C. D., Franklin, J. F., Frelich, L. E., Harvey, B. J., Higuera, P. E., ... Turner,
768 M. G. (2016). Changing disturbance regimes, ecological memory, and forest resilience. *Frontiers in*
769 *Ecology and the Environment*, 14 (7), 369–378. <https://doi.org/10.1002/fee.1311>
- 770 61. Kanehisa, M., Sato, Y., Furumichi, M., Morishima, K., & Tanabe, M. (2019). New approach for
771 understanding genome variations in KEGG. *Nucleic Acids Research*, 47 (D1), D590–D595.
772 <https://doi.org/10.1093/nar/gky962>

- 773 62. Kattge, J., Bönisch, G., Díaz, S., Lavorel, S., Prentice, I. C., Leadley, P., Tautenhahn, S., Werner, G.
774 D. A., Aakala, T., Abedi, M., Acosta, A. T. R., Adamidis, G. C., Adamson, K., Aiba, M., Albert, C. H.,
775 Alcántara, J. M., Alcázar C, C., Aleixo, I., Ali, H., ... Wirth, C. (2020). TRY plant trait database –
776 enhanced coverage and open access. *Global Change Biology*, 26 (1), 119–188.
777 <https://doi.org/10.1111/gcb.14904>
- 778 63. Kettenring, K. M., Mercer, K. L., Reinhardt Adams, C., & Hines, J. (2014). Application of genetic
779 diversity-ecosystem function research to ecological restoration. *Journal of Applied Ecology*, 51(2),
780 339–348. <https://doi.org/10.1111/1365-2664.12202>
- 781 64. Koch, E. L., & Guillaume, F. (2020). Additive and mostly adaptive plastic responses of gene
782 expression to multiple stress in *Tribolium castaneum*. *PLOS Genetics*, 16 (5), e1008768.
783 <https://doi.org/10.1371/journal.pgen.1008768>
- 784 65. Kokko, H., Chaturvedi, A., Croll, D., Fischer, M. C., Guillaume, F., Karrenberg, S., ... Stapley, J.
785 (2017). Can Evolution Supply What Ecology Demands? *Trends in Ecology & Evolution*, 32 (3), 187–
786 197. <https://doi.org/10.1016/j.tree.2016.12.005>
- 787 66. Kotta, J., Vanhatalo, J., Jänes, H., Orav-Kotta, H., Rugiu, L., Jormalainen, V., ... Johannesson, K.
788 (2019). Integrating experimental and distribution data to predict future species patterns. *Scientific*
789 *Reports*, 9 (1), 1821. <https://doi.org/10.1038/s41598-018-38416-3>
- 790 67. Krause, S. M. B., Meima-Franke, M., Veraart, A. J., Ren, G., Ho, A., & Bodelier, P. L. E. (2018).
791 Environmental legacy contributes to the resilience of methane consumption in a laboratory microcosm
792 system. *Scientific Reports*, 8 (1), 8862. <https://doi.org/10.1038/s41598-018-27168-9>
- 793 68. Legrand, D., Cote, J., Fronhofer, E. A., Holt, R. D., Ronce, O., Schtickzelle, N., ... Clobert, J. (2017).
794 Eco-evolutionary dynamics in fragmented landscapes. *Ecography*, 40 (1), 9–25.
795 <https://doi.org/10.1111/ecog.02537>
- 796 69. Leivesley, J. A., Bussière, L. F., Pemberton, J. M., Pilkington, J. G., Wilson, K., & Hayward, A. D.
797 (2019). Survival costs of reproduction are mediated by parasite infection in wild Soay sheep. *Ecology*
798 *Letters*, 22 (8), ele.13275. <https://doi.org/10.1111/ele.13275>
- 799 70. Lever, J. J., van Nes, E. H., Scheffer, M., & Bascompte, J. (2014). The sudden collapse of pollinator
800 communities. *Ecology Letters*, 17 (3), 350–359. <https://doi.org/10.1111/ele.12236>

- 801 71. Levin, S. A. (1992). The problem of pattern and scale in ecology: The Robert H. MacArthur award
802 lecture. *Ecology*, 73 (6), 1943–1967.
- 803 72. Lewis, S. L., & Maslin, M. A. (2015). Defining the Anthropocene. *Nature*, 519 (7542), 171–180.
804 <https://doi.org/10.1038/nature14258>
- 805 73. López-Maury, L., Marguerat, S., & Bähler, J. (2008). Tuning gene expression to changing
806 environments: from rapid responses to evolutionary adaptation. *Nature Reviews Genetics*, 9 (8), 583–
807 593. <https://doi.org/10.1038/nrg2398>
- 808 74. Lyons, S. K., Amatangelo, K. L., Behrensmeyer, A. K., Bercovici, A., Blois, J. L., Davis, M., ... Gotelli,
809 N. J. (2016). Holocene shifts in the assembly of plant and animal communities implicate human
810 impacts. *Nature*, 529 (7584), 80–83. <https://doi.org/10.1038/nature16447>
- 811 75. Madin, J. S., Anderson, K. D., Andreasen, M. H., Bridge, T. C. L., Cairns, S. D., Connolly, S. R., ...
812 Baird, A. H. (2016). The Coral Trait Database, a curated database of trait information for coral
813 species from the global oceans. *Scientific Data*, 3 (1), 160017. <https://doi.org/10.1038/sdata.2016.17>
- 814 76. Maley, C. C., Galipeau, P. C., Finley, J. C., Wongsurawat, V. J., Li, X., Sanchez, C. A., ... Reid, B. J.
815 (2006). Genetic clonal diversity predicts progression to esophageal adenocarcinoma. *Nature*
816 *Genetics*, 38 (4), 468–473. <https://doi.org/10.1038/ng1768>
- 817 77. Mason, I. C., Qian, J., Adler, G. K., & Scheer, F. A. J. L. (2020). Impact of circadian disruption on
818 glucose metabolism: implications for type 2 diabetes. *Diabetologia*, 63 (3), 462–472.
819 <https://doi.org/10.1007/s00125-019-05059-6>
- 820 78. Massot, M., Legendre, S., Féderici, P., & Clobert, J. (2017). Climate warming: a loss of variation in
821 populations can accompany reproductive shifts. *Ecology Letters*, 20 (9), 1140–1147.
822 <https://doi.org/10.1111/ele.12811>
- 823 79. McGill, B. J., Chase, J. M., Hortal, J., Overcast, I., Rominger, A. J., Rosindell, J., ... Gillespie, R.
824 (2019). Unifying macroecology and macroevolution to answer fundamental questions about
825 biodiversity. *Global Ecology and Biogeography*, 28 (12), 1925–1936.
826 <https://doi.org/10.1111/geb.13020>
- 827 80. McNamara, J. M., Dall, S. R. X., Hammerstein, P., & Leimar, O. (2016). Detection vs. selection:
828 integration of genetic, epigenetic and environmental cues in fluctuating environments. *Ecology*

829 *Letters*, 19 (10), 1267–1276. <https://doi.org/10.1111/ele.12663>

830 81. Meredith, H. R., Andreani, V., Ma, H. R., Lopatkin, A. J., Lee, A. J., Anderson, D. J., Batt, G., & You,
831 L. (2018). Applying ecological resistance and resilience to dissect bacterial antibiotic responses.
832 *Science Advances*, 4(12). <https://doi.org/10.1126/sciadv.aau1873>

833 82. Messier, C., Bauhus, J., Doyon, F., Maure, F., Sousa-Silva, R., Nolet, P., ... Puettmann, K. (2019).
834 The functional complex network approach to foster forest resilience to global changes. *Forest*
835 *Ecosystems*, 6 (1), 21. <https://doi.org/10.1186/s40663-019-0166-2>

836 83. Millar, C. I., Stephenson, N. L., & Stephens, S. L. (2007). Climate change and forests of the future:
837 managing in the face of uncertainty. *Ecological Applications*, 17 (8), 2145–2151.
838 <https://doi.org/10.1890/06-1715.1>

839 84. Nadeau, C. P., & Urban, M. C. (2019). Eco-evolution on the edge during climate change. *Ecography*,
840 42 (7), 1280–1297. <https://doi.org/10.1111/ecog.04404>

841 85. Oliver, T. H., Heard, M. S., Isaac, N. J. B., Roy, D. B., Procter, D., Eigenbrod, F., ... Bullock, J. M.
842 (2015). Biodiversity and resilience of ecosystem functions. *Trends in Ecology & Evolution*, 30 (11),
843 673–684. <https://doi.org/10.1016/j.tree.2015.08.009>

844 86. Oostra, V., Saastamoinen, M., Zwaan, B. J., & Wheat, C. W. (2018). Strong phenotypic plasticity
845 limits potential for evolutionary responses to climate change. *Nature Communications*, 9(1).
846 <https://doi.org/10.1038/s41467-018-03384-9>

847 87. Ovaskainen, O., & Abrego, N. (2020). *Joint Species distribution modeling with applications in R*.
848 Cambridge University Press.

849 88. Ovaskainen, O., Meyke, E., Lo, C., Tikhonov, G., Delgado, M. del M., Roslin, T., ... Kurhinen, J.
850 (2020). Chronicles of nature calendar, a long-term and large-scale multitaxon database on
851 phenology. *Scientific Data*, 7 (1), 47. <https://doi.org/10.1038/s41597-020-0376-z>

852 89. Palkovacs, E. P. & Hendry, A. P. (2010). Eco-evolutionary dynamics: Intertwining ecological and
853 evolutionary processes in contemporary time. *F1000 Biol. Rep.* 2, 1 (doi:10.3410/B2-1)

854 90. Papakostas, S., Vasemägi, A., Vähä, J.-P., Himberg, M., Peil, L., & Primmer, C. R. (2012). A
855 proteomics approach reveals divergent molecular responses to salinity in populations of European
856 whitefish (*Coregonus lavaretus*). *Molecular Ecology*, 21 (14), 3516–3530.

- 857 <https://doi.org/10.1111/j.1365-294X.2012.05553.x>
- 858 91. Papakostas, S., Vøllestad, L. A., Bruneaux, M., Aykanat, T., Vanoverbeke, J., Ning, M., ... Leder, E.
859 H. (2014). Gene pleiotropy constrains gene expression changes in fish adapted to different thermal
860 conditions. *Nature Communications*, 5 (1), 4071. <https://doi.org/10.1038/ncomms5071>
- 861 92. Pazzaglia, J., Reusch, T. B. H., Terlizzi, A., Marín-Guirao, L., & Procaccini, G. (2021). Phenotypic
862 plasticity under rapid global changes: The intrinsic force for future seagrasses survival. *Evolutionary*
863 *Applications*, 14 (5), 1181–1201. <https://doi.org/10.1111/eva.13212>
- 864 93. Pimm, S. L. (1984). The complexity and stability of ecosystems. *Nature*, 307 (5949), 321–326.
865 <https://doi.org/10.1038/307321a0>
- 866 94. Pontes-da-Silva, E., Magnusson, W. E., Sinervo, B., Caetano, G. H., Miles, D. B., Colli, G. R., ...
867 Werneck, F. P. (2018). Extinction risks forced by climatic change and intraspecific variation in the
868 thermal physiology of a tropical lizard. *Journal of Thermal Biology*, 73, 50–60.
869 <https://doi.org/10.1016/j.jtherbio.2018.01.013>
- 870 95. Primmer, C. R., Papakostas, S., Leder, E. H., Davis, M. J., & Ragan, M. A. (2013). Annotated genes
871 and nonannotated genomes: cross-species use of Gene Ontology in ecology and evolution research.
872 *Molecular Ecology*, 22 (12), 3216–3241. <https://doi.org/10.1111/mec.12309>
- 873 96. Roff G. (2021). Evolutionary history drives biogeographic patterns of coral reef resilience. *Bioscience*,
874 71 (1), 26–39. <https://doi.org/10.1093/biosci/biaa145/6030118>
- 875 97. Rudgers, J. A., Afkhami, M. E., Bell-Dereske, L., Chung, Y. A., Crawford, K. M., Kivlin, S. N., Mann,
876 M. A., & Nuñez, M. A. (2020). Climate disruption of plant-microbe interactions. *Annual Review of*
877 *Ecology, Evolution, and Systematics*, 51(1), 561–586. [https://doi.org/10.1146/annurev-ecolsys-](https://doi.org/10.1146/annurev-ecolsys-011720-090819)
878 011720-090819
- 879 98. Scheffer, M., Carpenter, S., Foley, J. A., Folke, C., & Walker, B. (2001). Catastrophic shifts in
880 ecosystems. *Nature*, 413 (6856), 591–596. <https://doi.org/10.1038/35098000>
- 881 99. Scheffer, M., Carpenter, S. R., Lenton, T. M., Bascompte, J., Brock, W., Dakos, V., ... Vandermeer, J.
882 (2012). Anticipating Critical Transitions. *Science*, 338 (6105), 344–348.
883 <https://doi.org/10.1126/science.1225244>
- 884 100. Schoener, T. W. (2011). The newest synthesis: understanding the interplay of evolutionary and

- 885 ecological dynamics. *Science*, 331 (6016), 426–429. <https://doi.org/10.1126/science.1193954>
- 886 101. Sgrò, C. M., Lowe, A. J., & Hoffmann, A. A. (2011). Building evolutionary resilience for conserving
887 biodiversity under climate change. *Evolutionary Applications*, 4 (2), 326–337.
888 <https://doi.org/10.1111/j.1752-4571.2010.00157.x>
- 889 102. Simberloff, D., Martin, J.-L., Genovesi, P., Maris, V., Wardle, D. A., Aronson, J., ... Vilà, M.
890 (2013). Impacts of biological invasions: what's what and the way forward. *Trends in Ecology &*
891 *Evolution*, 28 (1), 58–66. <https://doi.org/10.1016/j.tree.2012.07.013>
- 892 103. Standish, R. J., Hobbs, R. J., Mayfield, M. M., Bestelmeyer, B. T., Suding, K. N., Battaglia, L. L.,
893 ... Thomas, P. A. (2014). Resilience in ecology: Abstraction, distraction, or where the action is?
894 *Biological Conservation*, 177, 43–51. <https://doi.org/10.1016/j.biocon.2014.06.008>
- 895 104. Stange, M., Barrett, R. D. H., & Hendry, A. P. (2020). The importance of genomic variation for
896 biodiversity, ecosystems and people. *Nature Reviews Genetics*. [https://doi.org/10.1038/s41576-020-](https://doi.org/10.1038/s41576-020-00288-7)
897 [00288-7](https://doi.org/10.1038/s41576-020-00288-7)
- 898 105. Steane, D. A., Potts, B. M., McLean, E., Prober, S. M., Stock, W. D., Vaillancourt, R. E., & Byrne,
899 M. (2014). Genome-wide scans detect adaptation to aridity in a widespread forest tree species.
900 *Molecular Ecology*, 23 (10), 2500–2513. <https://doi.org/10.1111/mec.12751>
- 901 106. Strayer, D. L. (2010). Alien species in fresh waters: ecological effects, interactions with other
902 stressors, and prospects for the future. *Freshwater Biology*, 55 (s1), 152–174.
903 <https://doi.org/10.1111/j.1365-2427.2009.02380.x>
- 904 107. Strona, G., Lafferty, K. D., Fattorini, S., Beck, P. S. A., Guilhaumon, F., Arrigoni, R., Montano, S.,
905 Seveso, D., Galli, P., Planes, S., Parravicini, V. (2021). Global tropical reef fish richness could decline
906 by around half if corals are lost. *Proceedings of the Royal Society of London B*, 288
907 (1953), 20210274.
- 908 108. Sullivan, A. P., Bird, D. W., & Perry, G. H. (2017). Human behaviour as a long-term ecological
909 driver of non-human evolution. *Nature Ecology & Evolution*, 1 (3), 0065.
910 <https://doi.org/10.1038/s41559-016-0065>
- 911 109. Sun, M., Folk, R. A., Gitzendanner, M. A., Soltis, P. S., Chen, Z., Soltis, D. E., & Guralnick, R. P.
912 (2020). Recent accelerated diversification in rosids occurred outside the tropics. *Nature*

- 913 *Communications*, 11 (1), 1–12. <https://doi.org/10.1038/s41467-020-17116-5>
- 914 110. Sun, S.-J., & Kilner, R. M. (2020). Temperature stress induces mites to help their carrion beetle
915 hosts by eliminating rival blowflies. *ELife*, 9, e55649. <https://doi.org/10.7554/eLife.55649>
- 916 111. Sutherland, W. J., Freckleton, R. P., Godfray, H. C. J., Beissinger, S. R., Benton, T., Cameron, D.
917 D., ... Wiegand, T. (2013). Identification of 100 fundamental ecological questions. *Journal of Ecology*,
918 101, 58–67. <https://doi.org/10.1111/1365-2745.12025>
- 919 112. Swenson, N. G. (2019). *Phylogenetic ecology: A history, critique, and remodeling*. Chicago,
920 Illinois: University of Chicago Press.
- 921 113. Swinfield, T., Both, S., Riutta, T., Bongalov, B., Elias, D., Majalap-Lee, N., ... Coomes, D. (2020).
922 Imaging spectroscopy reveals the effects of topography and logging on the leaf chemistry of tropical
923 forest canopy trees. *Global Change Biology*, 26 (2), 989–1002. <https://doi.org/10.1111/gcb.14903>
- 924 114. The NOW Community. (2020). NOW — New and Old Worlds: Database of fossil mammals.
925 *Zenodo*, <http://doi.org/10.5281/zenodo.4268068>. <https://doi.org/10.5281/zenodo.4268068>
- 926 115. Thebault, E., & Fontaine, C. (2010). Stability of Ecological Communities and the Architecture of
927 Mutualistic and Trophic Networks. *Science*, 329 (5993), 853–856.
928 <https://doi.org/10.1126/science.1188321>
- 929 116. Thom, D., Golivets, M., Edling, L., Meigs, G. W., Gourevitch, J. D., Sonter, L. J., ... Keeton, W. S.
930 (2019). The climate sensitivity of carbon, timber, and species richness covaries with forest age in
931 boreal–temperate North America. *Global Change Biology*, 25 (7), 2446–2458.
932 <https://doi.org/10.1111/gcb.14656>
- 933 117. Tonteri, A., Vasemägi, A., Lumme, J., & Primmer, C. R. (2010). Beyond MHC: signals of elevated
934 selection pressure on Atlantic salmon (*Salmo salar*) immune-relevant loci. *Molecular Ecology*, 19
935 (7), 1273–1282. <https://doi.org/10.1111/j.1365-294X.2010.04573.x>
- 936 118. Tylianakis, J. M., & Maia, L. F. (2020). The patchwork of evolutionary landscapes. *Nature Ecology
937 and Evolution*, 4 (5), 672–673. <https://doi.org/10.1038/s41559-020-1184-8>
- 938 119. Udrescu, S.-M., & Tegmark, M. (2020). AI Feynman: A physics-inspired method for symbolic
939 regression. *Science Advances*, 6 (16), eaay2631. <https://doi.org/10.1126/sciadv.aay2631>
- 940 120. USGS. (2020). USGS datasets.

- 941 121. Valladares, F., Matesanz, S., Guilhaumon, F., Araújo, M. B., Balaguer, L., Benito-Garzón, M., ...
942 Zavala, M. A. (2014). The effects of phenotypic plasticity and local adaptation on forecasts of species
943 range shifts under climate change. *Ecology Letters*, 17 (11), 1351–1364.
944 <https://doi.org/10.1111/ele.12348>
- 945 122. van Kleunen, M., Dawson, W., Essl, F., Pergl, J., Winter, M., Weber, E., ... Pyšek, P. (2015).
946 Global exchange and accumulation of non-native plants. *Nature*, 525 (7567), 100–103.
947 <https://doi.org/10.1038/nature14910>
- 948 123. van Oppen, M. J. H., Oliver, J. K., Putnam, H. M., Gates, R. D. (2015). Building coral reef
949 resilience through assisted evolution. *Proceedings of the National Academy of Sciences USA*, 112
950 (8), 2307–13. [http:// doi/10.1073/pnas.1422301112](http://doi/10.1073/pnas.1422301112)
- 951 124. Verhagen, I., Tomotani, B. M., Gienapp, P., & Visser, M. E. (2020). Temperature has a causal
952 and plastic effect on timing of breeding in a small songbird. *The Journal of Experimental Biology*, 223
953 (8), jeb218784. <https://doi.org/10.1242/jeb.218784>
- 954 125. Waldvogel, A., Feldmeyer, B., Rolshausen, G., Exposito-Alonso, M., Rellstab, C., Kofler, R.,
955 Mock, T., Schmid, K., Schmitt, I., Bataillon, T., Savolainen, O., Bergland, A., Flatt, T., Guillaume, F., &
956 Pfenninger, M. (2020). Evolutionary genomics can improve prediction of species' responses to
957 climate change. *Evolution Letters*, 4(1), 4–18. <https://doi.org/10.1002/evl3.154>
- 958 126. Waples, R. S., Beechie, T., & Pess, G. R. (2009). Evolutionary History, Habitat Disturbance
959 Regimes, and Anthropogenic Changes: What Do These Mean for Resilience of Pacific Salmon
960 Populations? *Ecology and Society*, 14 (1), art3. <https://doi.org/10.5751/ES-02626-140103>
- 961 127. Ware, I. M., Fitzpatrick, C. R., Senthilnathan, A., Bayliss, S. L. J., Beals, K. K., Mueller, L. O., ...
962 Bailey, J. K. (2019). Feedbacks link ecosystem ecology and evolution across spatial and temporal
963 scales: Empirical evidence and future directions. *Functional Ecology*, 33 (1), 31–42.
964 <https://doi.org/10.1111/1365-2435.13267>
- 965 128. Weise, H., Auge, H., Baessler, C., Bärlund, I., Bennett, E. M., Berger, U., ... Grimm, V. (2020).
966 Resilience trinity: safeguarding ecosystem functioning and services across three different time
967 horizons and decision contexts. *Oikos*, 129 (4), 445–456. <https://doi.org/10.1111/oik.07213>
- 968 129. Whitehead, H., Laland, K. N., Rendell, L., Thorogood, R., & Whiten, A. (2019). The reach of

969 gene–culture coevolution in animals. *Nature Communications*, 10 (1), 2405.
970 <https://doi.org/10.1038/s41467-019-10293-y>

971 130. Wilkinson, M. D., Dumontier, M., Aalbersberg, Ij. J., Appleton, G., Axton, M., Baak, A., ... Mons,
972 B. (2016). The FAIR guiding principles for scientific data management and stewardship. *Scientific*
973 *Data*, 3 (1), 160018. <https://doi.org/10.1038/sdata.2016.18>

974 131. Winter, M., Devictor, V., & Schweiger, O. (2013). Phylogenetic diversity and nature conservation:
975 where are we? *Trends in Ecology & Evolution*, 28 (4), 199–204.
976 <https://doi.org/10.1016/j.tree.2012.10.015>

977 132. Wong, B. B. M., & Candolin, U. (2014). Behavioral responses to changing environments.
978 *Behavioral Ecology*, 26, 665–673. <https://doi.org/10.1093/beheco/aru183>

979 133. Yang, Q., Fowler, M. S., Jackson, A. L., & Donohue, I. (2019). The predictability of ecological
980 stability in a noisy world. *Nature Ecology & Evolution*, 3 (2), 251–259. [https://doi.org/10.1038/s41559-](https://doi.org/10.1038/s41559-018-0794-x)
981 [018-0794-x](https://doi.org/10.1038/s41559-018-0794-x)

982 134. Zitnik, M., Sosič, R., Feldman, M. W., & Leskovec, J. (2019). Evolution of resilience in protein
983 interactomes across the tree of life. *Proceedings of the National Academy of Sciences*, 116 (10),
984 4426–4433. <https://doi.org/10.1073/pnas.1818013116>

985