

1 **Sources of confusion in global biodiversity trends**

2 **Authors:** Maelys Boennec^{1*}, Vasilis Dakos¹, Vincent Devictor¹

3 **Affiliations:**

4 ¹Institut des Sciences de l'Évolution de Montpellier, Montpellier, France (34000)

5 ***Correspondence to:**

6 Maelys Boennec (maelys.boennec@umontpellier.fr)

7

8 **Abstract**

9 Populations and ecological communities are changing worldwide, and empirical studies exhibit a
10 mixture of either declining or mixed trends. Confusion in global biodiversity trends thus remains
11 while being of major social, political, and scientific importance. Part of this variability may arise
12 from the difficulty to reliably assess global biodiversity trends. Here, we conducted a literature
13 review of studies documenting the temporal dynamics of global biodiversity. We classified the
14 differences among approaches, data and methodology used by the reviewed papers to reveal
15 common findings and sources of discrepancies. We show that reviews and meta-analyses, along
16 with the use of global indicators, are more likely to conclude that trends are declining. On the other
17 hand, the longer the data are available, the more nuanced are the trends they generate. Our results
18 also highlight the lack of studies providing information on the impact of synergistic pressures on a
19 global scale, making it even more difficult to understand the driving factors of the observed changes
20 and how to decide conservation plan accordingly. Finally, we stress the importance of taking into
21 account the sources of confusion identified, as well as the complexity of biodiversity changes, in
22 order to implement effective conservation strategies. In particular, biodiversity dynamics are almost
23 systematically assumed to be linear, while non-linear trends are largely neglected. Clarifying the
24 sources of confusion in global biodiversity trends should strengthen large scale biodiversity
25 monitoring and conservation.

26 **Introduction**

27 Providing a coherent synthesis of the ongoing biodiversity crisis through the quantification of
28 various aspects of temporal changes of biological diversity is a challenge of both scientific and
29 political importance. The accumulation of studies reporting a loss of biodiversity, either in species
30 number (IUCN, 2019), population abundances (WWF, 2022) or at the assemblage or community
31 scale (Mckinney & Lockwood, 1999; Olden et al., 2004; Sax & Gaines, 2003), leave no doubt
32 regarding the fact that biodiversity is being depleted. However, empirical reports of temporal
33 changes in biological diversity depict a nuanced and complex picture regarding the magnitude and
34 direction of biodiversity loss, encompassing findings that may intuitively be seen as opposite. In
35 addition to studies quantifying a global decline of biological diversity, others have suggested that
36 biodiversity is not, on average, in decline (Blowes et al., 2019; Dornelas et al., 2014, 2019; Vellend
37 et al., 2013). While highlighting a significant turnover in species composition, these analyses found
38 no evidence of a systematic decline in species richness. Other recent studies revealed that only a
39 few declining species were mostly responsible for the negative trends in overall indices (Leung et
40 al., 2020), and demonstrated a rather balanced number of increasing and decreasing population
41 trends worldwide (Daskalova et al., 2020; Dornelas et al., 2019).

42
43 These heterogeneities in results entails several risks if their sources and uncertainties are not
44 addressed. Firstly, it could lead to sub-optimal or, worse, ineffective conservation policies, as many
45 conservation measures rely on the estimation of indicators or discussion of scenarios covering
46 global biodiversity trend for all taxa on a global scale (Agardy, 2005; Pressey et al., 2007).
47 Furthermore, avoiding to question the sources of the heterogeneity of these results and the
48 uncertainties in the conclusions could encourage a "biodiversity-skepticism" by creating the idea
49 that there is a lack of scientific consensus on the existence of a biodiversity crisis. Similarly,
50 climate-skepticism partly emerged from the belief that there were significant disagreements about
51 global warming among scientists (Joslyn & LeClerc, 2016; Leiserowitz et al., 2013). Such distorted
52 perception was reduced by incorporating uncertainties into IPCC reports, clarifying the fact that
53 much of the variability was due to predictive processes rather than fundamental differences in
54 scientific opinion (Howe et al., 2019; Reilly et al., 2001). Thus, conservation science and
55 knowledge on biodiversity loss should also benefit from such clarification and be consolidated by
56 adopting a transparent and quantitative approach to major biases in the global estimates.

57 If these heterogeneous results have also caused a vivid controversy in the scientific community
58 (Cardinale, 2014; Cardinale et al., 2018; Gonzalez et al., 2016), it is to some extent probably
59 resulting from the multiple meanings of the term *biodiversity*. The formal definition largely

60 popularized by the Rio Earth Summit in 1992 equates biodiversity to "the variability among living
61 organisms from all sources [...] this includes diversity within species, between species and of
62 ecosystems" (CBD, 1992). This definition itself creates a lot of confusion. For example, the WWF
63 considers biodiversity at the population level (WWF, 2020, 2022). In contrast, others (e.g. Dornelas
64 et al., 2014) consider biodiversity at the species or community level. As declines in population sizes
65 and species richness are not necessarily related, both an increase and decrease in "biodiversity" can
66 be concluded depending on the ecological level of interest. Clarifying the trends for each level of
67 biodiversity should, in principle, limit the confusion. In practice however, trends within the same
68 ecological levels show both a decline globally or no net changes; either at species scale (IUCN,
69 2019; Dornelas et al., 2014; Vellend et al., 2013) or at population scale (Daskalova et al., 2020; He
70 et al., 2019; Leung et al., 2020; Wagner et al., 2021). Beyond the ecological level considered other
71 factors are therefore also generating confusion in the observed results.

72
73 The same difficulties affect the understanding of which and how pressures drive temporal changes
74 of biodiversity on a global scale. The drivers of biodiversity loss are widely documented (Ceballos
75 et al., 2015; IPBES, 2019; Pereira et al., 2012; Pievani, 2014; Pimm et al., 2006). The effects of
76 climate change and of anthropogenic drivers – e.g. habitat fragmentation – have been studied at the
77 individual level – e.g. through changes in physiology (Willis & Bhagwat, 2009) –, at the species
78 and population levels, or at the community level – e.g. through changes in interspecific
79 relationships (Gilman et al., 2010; Walther, 2010) –, or either spatially – e.g. through range shifts
80 (Erauskin-Extramiana et al., 2020; Paprocki et al., 2014) –, or temporally – e.g. through changes in
81 phenology (Du et al., 2019; Radchuk et al., 2019; Wolf et al., 2017). However, the responses of
82 different ecological levels to specific drivers are mostly documented at the local scale. The
83 understanding of how global change drivers influence the heterogeneous biodiversity patterns at the
84 global scale is therefore also limited, and filled with controversies. For instance, some studies
85 suggest that habitat fragmentation may be beneficial to biodiversity (Fahrig, 2017; Haddad et al.,
86 2017) or that protected areas often fail to reduce biodiversity loss (Brooke et al., 2008; Mora &
87 Sale, 2011) and can even be detrimental to biodiversity (Geldmann et al., 2019).

88
89 Acknowledging our current sources of (mis)understanding of the temporal changes of biodiversity
90 and of their drivers at the global scale is urgently needed (Tekwa et al., 2023). International
91 legislation and objectives (such as those discussed at United Nations Biodiversity Conference of the
92 Parties) directly rely on our general knowledge and understanding of global biodiversity dynamics.
93 The objective of this study is therefore to critically examine this general knowledge. More

94 specifically, we want to review how global biodiversity change is currently quantified, and to
95 identify the most salient sources of confusion when assessing biodiversity trends or the effect of its
96 drivers.

97 Several hypotheses can be formulated. First, with regard to trends, the data used can be expected to
98 affect the results. The prevalence of certain threatened groups (Houlahan et al., 2000; Marsh, 2001),
99 the lack of spatial representation (e.g. tropics are highly biodiverse (Laurance, 2007) but lack data
100 representation (Feeley & Silman, 2011)), or the temporal extent of the time series used (Duchenne
101 et al., 2022; Gonzalez et al., 2016; Vellend et al., 2013) are likely to impact the conclusions. We also
102 hypothesise that the methods with which trends are quantified play a role. Several studies have
103 already pointed out that summarizing complex data using global indicators can hide meaningful
104 variation (Daskalova et al., 2020; Leung et al., 2020), but it is not clear to what extent and whether
105 this is the case for other methods. Finally, different approaches, from meta-analyses and reviews of
106 local studies to the analysis of globally aggregated empirical data, might also affect the conclusions.
107 Regarding the influence of drivers, we also expect that the biodiversity data used impact the results.
108 Focusing research on specific areas of the world can bias our knowledge. For example, the Arctic is
109 projected to warm at two up to four times the rate of the global average but it is understudied
110 (IPCC, 2021). A patchy data collection across taxa (or periods) also risks missing particularly
111 (un)sensitive groups (or (un)stable periods) (Mihoub et al., 2017). Variations are also expected
112 depending on the approach, the methods and the drivers considered.
113 Here, we conduct a literature review to clarify those sources of confusion, and to review the
114 remaining challenges for future research and conservation science.

115

116 **Methods**

117

118 ***Literature review***

119 We conducted a literature review in February 2022 aiming to identify papers providing an
120 assessment of recent temporal changes of biodiversity globally. We were looking both for papers
121 that studied the changes themselves, but also those that sought to explain the changes by studying
122 the impact of the drivers. Different searches were initially launched in the *Web of Science*. We tested
123 different search terms to refine the results. We wanted to minimise the number of irrelevant
124 references while ensuring that some commonly known relevant articles (e.g. Dornelas et al., 2019)
125 fell within our scope. In the process, we identified broad terms (e.g. "population") that encompassed
126 areas of research beyond our topic. We also found that restricting the search to terms in the titles
127 and abstracts, rather than in the topics of the articles in general, allowed us to limit the search to a

128 manageable number of articles (Appendix S1). The final search we launched in the *Web of Science*
129 was thus: TI=((biodiversity OR population* OR communit* OR indicator* OR natur* OR richness
130 OR species OR "biological diversity" OR abundance OR assemblage OR *flora OR *fauna) AND
131 (trend* OR dynamic* OR "time series" OR declin* OR loss OR extinct* OR increas* OR gain OR
132 coloni* OR change* OR fluctuat* OR trajector* OR tempo*)) AND AB=(("temporal" OR time)
133 AND ("analys*" OR "model*" OR stud* OR quantifi*)) NOT TI=("human population" OR "urban
134 population") AND (TI=(global OR worldwide) OR AB=(global OR worldwide)) NOT
135 WC=("meteorology atmospheric sciences" OR "infectious disease" OR "biochemistry molecular
136 biology" OR "paleontology" OR "microbiology"). TI is title, AB is abstract and WC is Web of
137 Science Categories. This query resulted in 2,008 matches.

138

139 Each of these papers was then reviewed individually by titles, abstracts and then subjected to a full
140 review to check their relevance to our study objective. We included studies that either assess or
141 discuss the assessment of (i) temporal changes of biodiversity ; (ii) during the last century ; (iii) on
142 a broad scale (at least two continents or two oceans). We included studies analysing temporal
143 changes on biodiversity relying on empirical data, but also reviews based on empirical assessments
144 as well as methodological studies explicitly questioning the assessment of global biodiversity
145 changes. For studies relying on data, we excluded those using exclusively human-manipulated or
146 simulated data. Figure 1A summaries this selection process and shows the number of articles
147 excluded on the basis of their failure to meet the above criteria. In addition, we included four
148 reports from the ‘Grey literature’ (IPBES, 2019; *IUCN*, 2022; Pörtner et al., 2021; WWF, 2020),
149 produced by organisations among the best known that provide assessments of temporal changes of
150 biodiversity globally. 91 references constituted the final database, that we classified into four
151 different categories: (i) biodiversity empirical analysis (n=48) ; (ii) reviews (n=20) ; (iii)
152 methodological papers (n=19) ; (iv) reports (n=4).

153

154 ***Metadata extraction***

155 Methodological papers were considered for discussion purposes. For the other references, we
156 extracted detailed metadata and information to investigate potential sources of heterogeneities
157 regarding the Global Biodiversity Changes (Fig. 1B). We distinguished papers mostly focused on
158 biodiversity trends from those mostly focused on drivers. Papers that were studying both trends and
159 drivers were considered both for analysing trends but also for analysing drivers, as we analyzed
160 these issues separately (n=6).

161

162 For papers analyzing biodiversity trends, we recorded bibliometric data, the main conclusions, the
163 type of assessment approach, and, when relevant, information regarding the data and methods used
164 to quantify the changes (Fig. 1B). Conclusions were classified into decreasing trends, increasing
165 trends, mixed trends (i.e. there were as many increasing trends as decreasing, or a majority of no
166 trends) and factor-dependent trends (i.e. directions that varied according to certain factors, such as
167 location).

168

169 While retrieving those information, we identified two main assessment approaches adopted in the
170 corresponding studies to produce a global picture of biodiversity changes. First, a bottom-up
171 approach, which correspond to reviews or meta-analyses aggregating results of studies analyzing
172 individual datasets with varying methodologies from one study to the other (e.g. Pereira et al., 2012;
173 Pievani, 2014). Second, a top-down approach that produces a global result analyzing heterogeneous
174 data aggregated in large databases within the same methodological framework (e.g. Dornelas et al.,
175 2019; Wilson & Fox, 2021). We recorded the assessment approach for each reference and decided
176 to investigate to what extent these could explain part of the observed heterogeneous results.

177

178 When relevant and available, we also retrieved information on the data used, the ecological level
179 targeted and the methodological processes used to quantify the changes. We recorded the databases
180 used, temporal and spatial scope, taxonomic scope, and number of species.

181 There are many measures of biodiversity, but in the papers selected we found measures of
182 biodiversity at the population scale (abundance, density or biomass), at the species scale (mostly
183 species richness), and at the community scale (assemblage composition, studied through similarity
184 indices).

185 We categorised the methods used to quantify the changes into three main categories. The first
186 category we identified concerns papers that studied the changes through linear trends of individual
187 ecological level. These papers performed linear regressions (and all variations, e.g. state space
188 models (Daskalova et al., 2020) or logged annual growth rate (Williams et al., 2022)) on
189 biodiversity measures either population by population or species by species. These results were then
190 used to reach conclusions about the general direction of change. We will refer to these methods as
191 “linear models on individual time series”. The second category we identified corresponds to papers
192 producing global indicators built upon populations (e.g. Living Planet Index (LPI) (Loh et al.,
193 2005)) or species (e.g. Red List Index (RLI)), and estimating global trends of those single
194 aggregated metrics. The last category we identified are the other methods that were used more
195 sporadically, quantifying other aspects than the linear changes in abundance or species richness (e.g.

196 quantifying the coefficient of variation (Marsh, 2001), non-linear changes (Keith et al., 2015), or
197 other aspects like resistance and recovery (Capdevila et al., 2022)).

198

199 For papers analyzing drivers of biodiversity changes, we recorded bibliometric data, the main
200 conclusions, the type of assessment approach, and, when relevant, information regarding the
201 biodiversity data used and the identity of the studied drivers (Fig. 1B). Conclusions were classified
202 into different categories based on the nature of the impacts into negative, none or positive, and
203 factor-dependent. The assessment approaches and information regarding the biodiversity data were
204 the same as the ones described regarding the examination of the trends in biodiversity. The
205 information provided on the drivers and the methods used to quantify the drivers' impacts on
206 biodiversity was scarce and heterogeneous. As a result, we did not address the issue of the
207 sensitivity of the results to the drivers data and methods used in depth. We rather examined the
208 impact of the identity of the drivers studied. We declined the drivers between climate change, other
209 anthropogenic pressures (mainly land use change) and conservation policies (protected areas).

210

211 **Results**

212

213 ***Global overview***

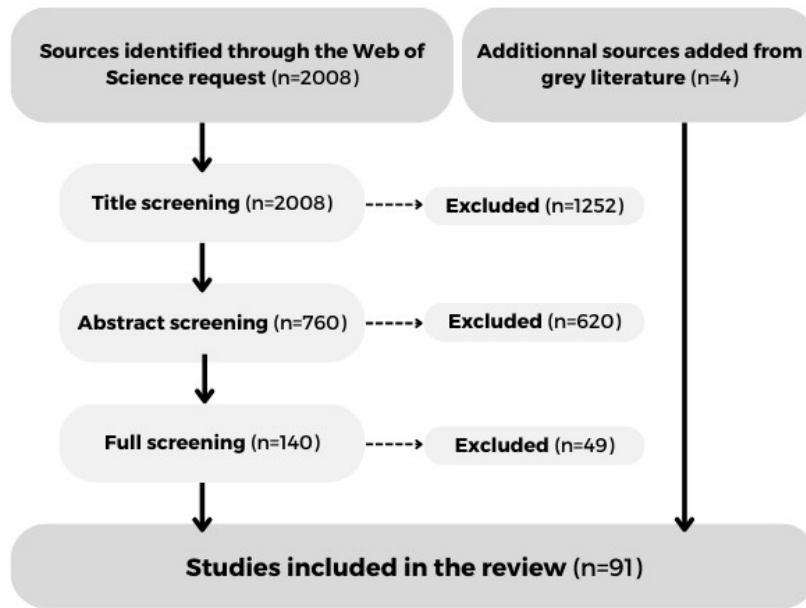
214 Interestingly, we found only 91 papers that analyze temporal changes of biodiversity globally, while
215 a plethora of examples accumulate at more local scales (e.g. Donald et al., 2001; Koleček et al.,
216 2021). These were published from 1991 to 2021, with a majority published after 2010 (Appendix
217 S2). We identified an increasing interest in this question over time, although this tendency also
218 reflects an overall increase in the number of papers published during the same period. Of the 91
219 papers, 48 directly relied on empirical datasets, 20 were reviews, 19 were methodological papers
220 and 4 were reports that we added from grey literature.

221

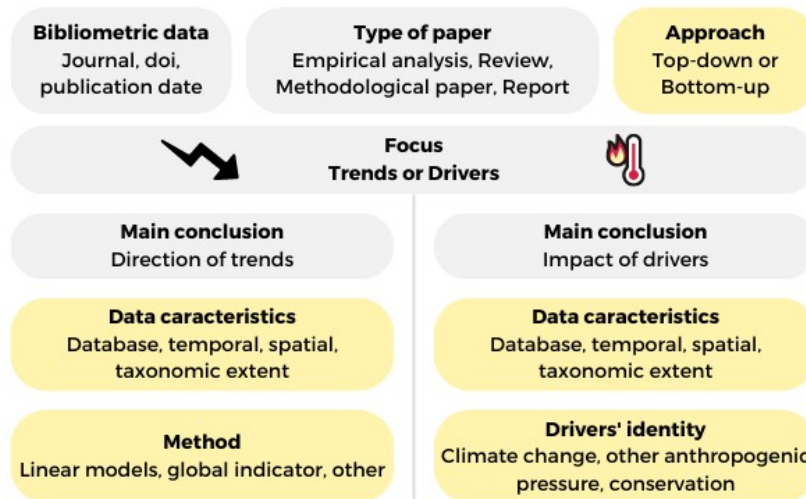
222 Out of the 44 papers assessing temporal trends in biodiversity changes, 57% (n=25) concluded that
223 biodiversity is globally in decline. More than a third of the papers concluded that trends are mixed
224 or factor-dependent (36%, n=16). Only 3 of the reviewed papers showed evidence of increasing
225 biodiversity trends. Out of the 34 studies focusing on drivers, 65% (n=22) concluded negative
226 effects on biodiversity, 26% (n=9) concluded that the effects were factor-dependent, and only 3
227 papers concluded that there was no significant effect or that the effect was positive (the latter
228 referring to the effect of conservation plans only).

229

A Screening process



B Metadata extraction



230 **Figure 1. Summary of the sampled literature and extracted metadata.** (A) Diagram
 231 representing the different steps of the screening process. List of studies extracted from the *Web of*
 232 *Science*, including excluded studies with reasons for exclusion are in the Appendix S3. (B)
 233 Summary of the metadata collected for the database. The potential sources of heterogeneities we
 234 investigate are highlighted in yellow.

235 **Table 1. Overview of the different sources of data identified.**

Data source	Biotime ($n_{tot}=7$)	Living Planet Database (LPD) ($n_{tot}=8$)	Global Population Dynamics Database (GPDD) ($n_{tot}=4$)	Other* ($n_{tot}=5$)	Aggregation** ($n_{tot}=30$)
Content	Time series data of ecological assemblages (mainly used at community <i>and</i> species level)	Time series data of individual species' abundance (mainly used at population level)	Time series data of individual species' abundance (mainly used at population level)	Time series data of several biodiversity measures (both species and population level)	Time series data of several biodiversity measures (mainly used at species <i>or</i> population level)
Taxa	Vertebrates, invertebrates and plants (8 in total)	Vertebrates (5 in total)	Vertebrates, invertebrates and plants (8 in total)	3 on average	3 on average
Realms	Marine, terrestrial and freshwater	Marine, terrestrial and freshwater	Marine, terrestrial and freshwater	Marine and terrestrial	Marine, terrestrial and freshwater
Average number of species	13,726 ($n_{calc}=6$)	847($n_{calc}=8$)	424 ($n_{calc}=3$)	30,553 ($n_{calc}=3$)	2,905 ($n_{calc}=21$)

236 Description of the five types of encountered databases providing time series data on different
 237 biodiversity levels. n_{tot} represents the total number of articles considering each of the data source.
 238 n_{calc} represents the number of articles used to calculate the averages (without those whose
 239 information is unavailable).

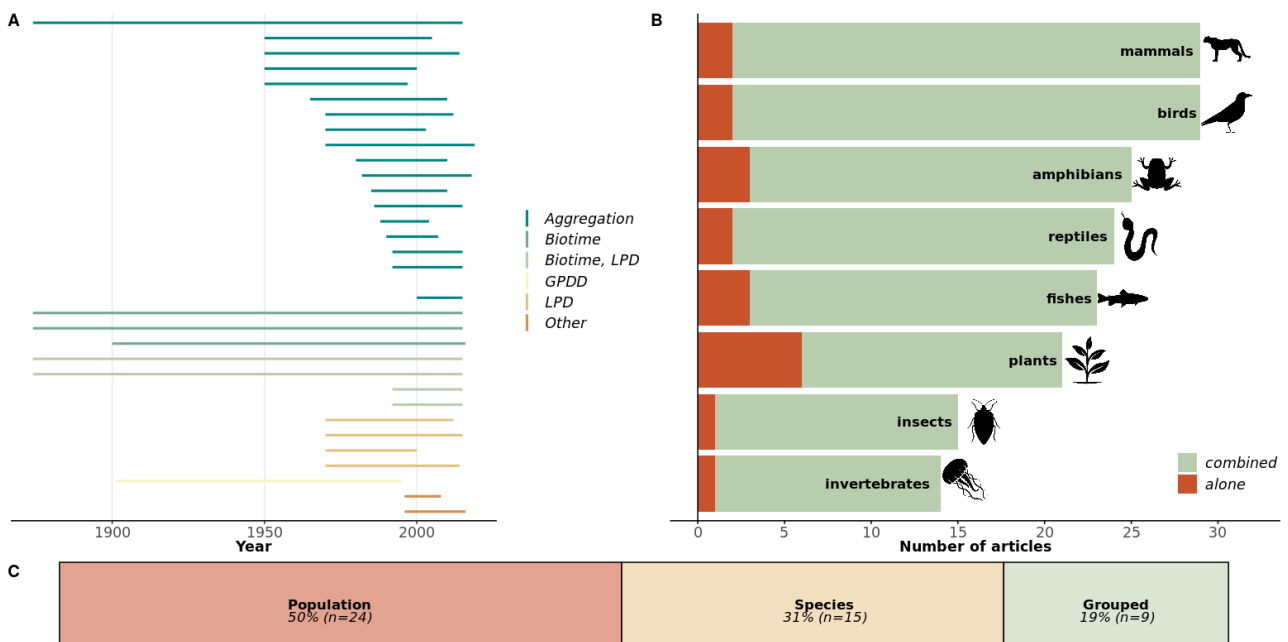
240 * Single non-aggregated but marginal data sets (used in articles only once).

241 ** Aggregation of several locally available data sets in individual studies.

242

243 Papers relying on empirical data used in majority aggregations of several data sets from individual
 244 studies (62,5%, $n=30$). Still, we found that three main global databases gathering biodiversity data
 245 were used: BioTIME (Dornelas et al., 2018), the Living Planet Database (LPD) (Loh et al., 2005),
 246 and the Global Population Dynamics Database (GPDD) (Inchausti & Halley, 2001). The remaining
 247 papers relied on other unique but more sporadically used databases that were not open access. Table
 248 1 highlights the characteristics and properties of the different data sources. The overwhelming
 249 majority of papers were based on terrestrial data covering all continents (67%, $n=32$), which is not
 250 surprising since our selection process was based in part on the choice to have global studies. Still,
 251 we accepted articles covering at least two continents, and we note that the only continents that were
 252 studied in pair were America and Europe. The temporal coverage of the data was quite

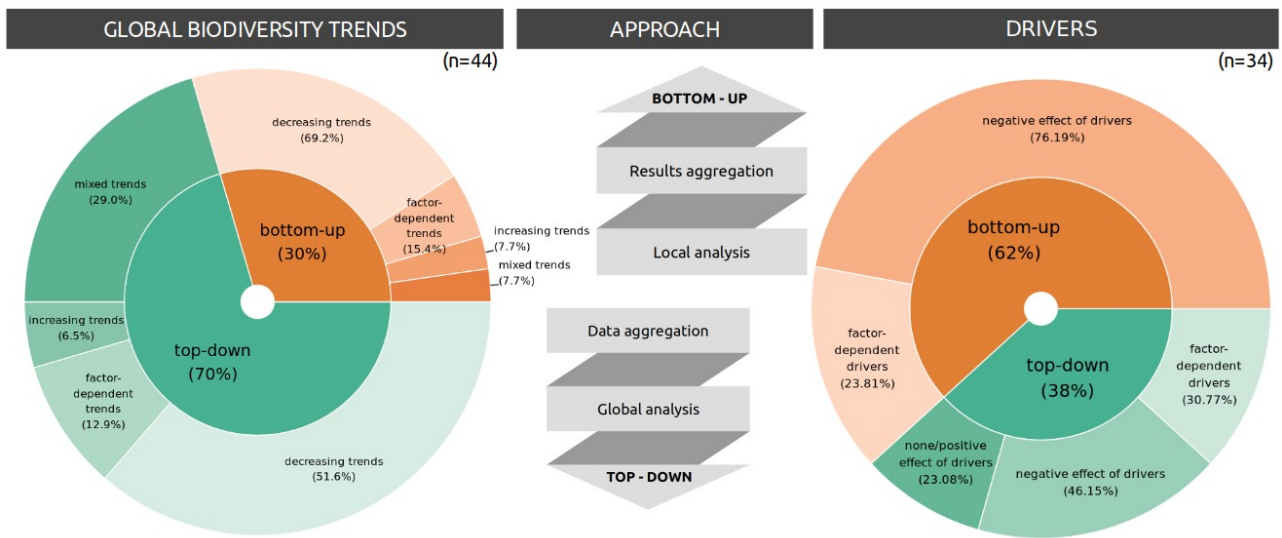
253 heterogeneous, but mainly covered the last 70 years (Fig. 2A). This also confirms the strength of
 254 BioTIME's temporal coverage compared to other databases (Dornelas et al., 2018). The taxonomic
 255 coverage shows that the papers in our corpus mainly relied on data referring to several groups (Fig.
 256 2B). As well as for the geographical extent, this is not surprising considering our selection process.
 257 Still, there is a preponderance of data on mammals and birds, more generally on terrestrial
 258 vertebrates, and an under-representation of insects and invertebrates.
 259 The empirical data were based on different levels of biodiversity (Fig. 2C). Papers relying on
 260 biodiversity at the population scale (e.g. abundances, biomass, density) represented 50% of our
 261 corpus (n=24), papers relying on biodiversity at the species scale (e.g. species richness, detection
 262 rates) represented 31% (n=15). The 19% remaining relied on grouped levels, i.e. several levels
 263 among population, species, or also community scale (studied through community composition).
 264



265 **Figure 2. Information on the biodiversity data on which the reviewed papers are based.** (A)
 266 Temporal extent of the data, colored based on the data source (see Table 1). (B) Taxonomic extent
 267 of the data. Number of papers considering each group, whether considered alone or combined (i.e.
 268 when the study does consider several groups at a time). For instance, 2 papers were examining the
 269 fate of mammals only and 30 were examining the fate of mammals and other groups at the same
 270 time. (C) Ecological level representation. Number of papers and percentages are indicated.
 271 « Grouped » category represents papers examining several ecological levels within the same paper.

272 **Potential sources of heterogeneities in the assessment of biodiversity trends**

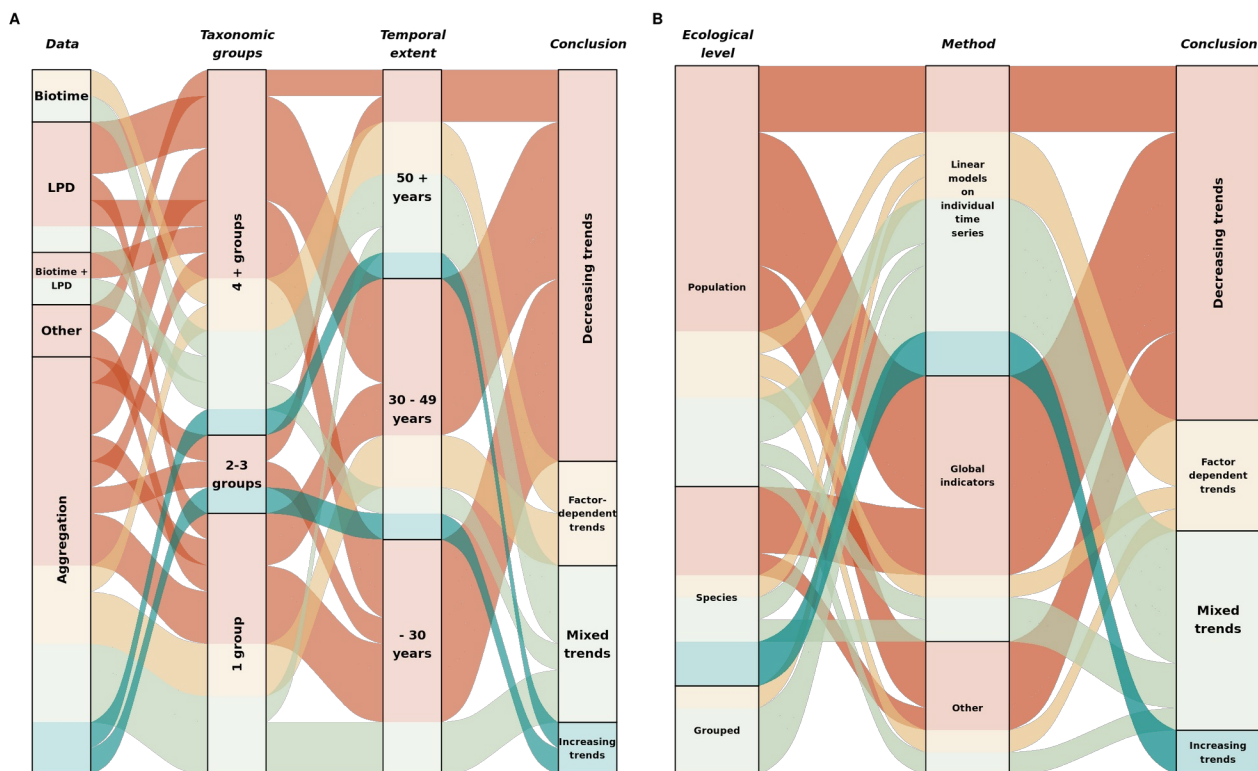
273 We investigated three potential sources of heterogeneities. First, we examined the effect of using
 274 different assessment approaches. When papers used a bottom-up approach (i.e. aggregating local or
 275 regional results with varying methodologies into meta-analyses or reviews), they mostly concluded
 276 a decline in global biodiversity (69%, Fig. 3). In contrast, when papers used a top-down approach
 277 (i.e. performing analyses through a single methodology on global datasets), their conclusions were
 278 much more balanced: a bit more than 50% concluded that biodiversity was decreasing and almost
 279 50% that trends were mixed or increasing (Fig. 3).



280 **Figure 3. Proportions within conclusions depending on whether global biodiversity trends or**
 281 **drivers were assessed and depending on the assessment approach (i.e. top-down or bottom-**
 282 **up).**

284 Second, we examined the effect of using data with different characteristics. Among the 22 papers
 285 using aggregated databases, 45% concluded a decline, 41% mixed or factor-dependent trends and
 286 14% identified increases in biodiversity globally. The LPD and the other individual datasets were
 287 mainly associated with a decline in global biodiversity, whereas studies using the BioTIME
 288 database mainly concluded that trends were mixed or factor-dependent (Fig. 4A). The conclusions
 289 between declining or mixed trends were balanced no matter the number of taxonomic groups
 290 considered (Fig. 4A). The time span did influence the conclusions however. The longer the data, the
 291 more heterogeneous the results were: 78% of articles based on time series of less than 30 years
 292 concluded that trends were declining, whereas this percentage dropped to 60% for time series
 293 between 30 and 49 years long, and to 22% for time series longer than 50 years.

294



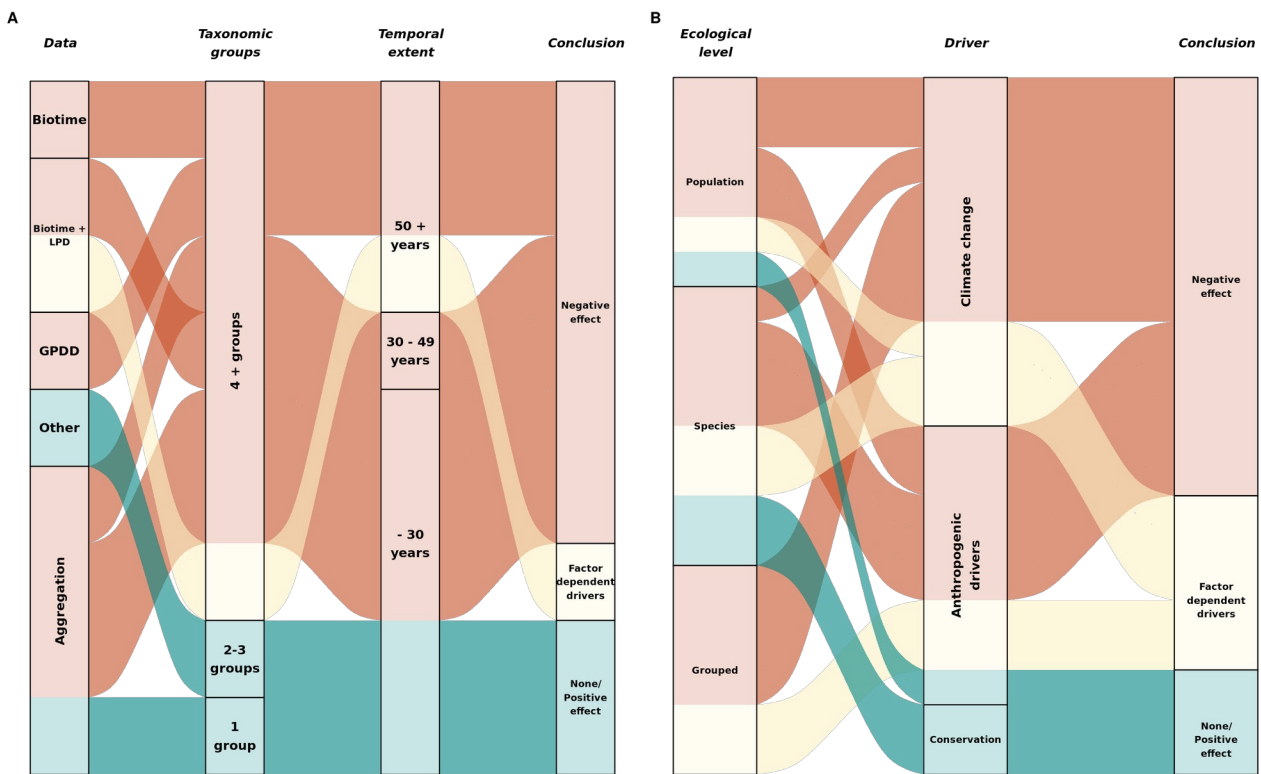
295 **Figure 4. Alluvial plots representing the links between the potential sources of heterogeneity**
 296 **and the conclusions when assessing trends.** A (n= 27) highlights the impact of the data
 297 characteristics. B (n=32) highlights the impact of the methods and the ecological level. The height
 298 of the boxes is proportional to the number of papers. Flow thickness between boxes is proportional
 299 to the number of papers. Colours of the flow reflect the conclusions.

300

301 Third, we examined the effect of using different methods. Among the 12 papers we reviewed that
 302 used global indicators to assess biodiversity changes, 9 concluded that biodiversity was declining,
 303 and only 3 that trends were more mixed. On the other hand, papers using linear models on
 304 individual time series depicted a much more nuanced picture, with 9 papers concluding that trends
 305 were mixed, 3 that trends were declining and 2 that trends were increasing. The other methods
 306 weren't much represented but concluded declines in 4 papers out of 6. Surprisingly, papers focusing
 307 on the population level showed more declines than the ones focusing on the species level (67 %
 308 declines against 44 % declines respectively). The papers relying on several levels of biodiversity
 309 only concluded that trends were mixed. However, as Figure 4B illustrates, these conclusions seem
 310 to be highly driven by the underlying quantification methods. Indeed, declines identified at both
 311 species or population levels are in majority declines that are identified through the use of global
 312 indicators.

313 **Potential sources of heterogeneities in the assessment of biodiversity drivers**

314 We also explored the effect of three potential sources of heterogeneities when assessing biodiversity
 315 drivers' impacts. First, we examined the effect of using different assessment approaches. Among the
 316 papers using a top-down approach, there was an almost even split between concluding a negative
 317 effect of the drivers considered on biodiversity (46%) and a factor-dependent or no effect (54%). In
 318 contrast, 76% of the papers using a bottom-up approach concluded that the drivers identified had a
 319 negative impact (Fig. 5).



320 **Figure 5. Alluvial plots representing the links between the potential sources of heterogeneity**
 321 **and the conclusions when assessing the impact of the drivers.** A (n=9) highlights the impact of
 322 the data characteristics. B (n=20) highlights the impact of the biodiversity level and the identity of
 323 the drivers. The height of the boxes is proportional to the number of papers. Flow thickness between
 324 boxes is proportional to the number of papers. Colours of the flow reflect the conclusions.
 325

326 The second source of heterogeneity we explored is related to the characteristics of biodiversity data
 327 that were used. Patterns regarding the impact of drivers did not clearly show a data bias (Fig. 5A),
 328 probably because of the very low number of papers for which enough information was available
 329 (n=9). Only papers using single and rarely used databases or aggregated databases (associated with
 330 poor taxonomic and temporal coverage) concluded that drivers did not have an effect on observed
 331 changes (n=2). The majority of papers (n=6), regardless of the data used, concluded that drivers had
 332 a negative impact on biodiversity.

333 The third source of heterogeneity we tested was the identity of the assessed drivers. We highlighted
334 that only conservation measures had a positive effect on biodiversity. Climate change affected
335 biodiversity mostly negatively, but conclusions were more nuanced concerning other anthropogenic
336 drivers (Fig. 5B).

337

338 **Discussion**

339 Quantification of temporal changes in biodiversity at large scales and attribution of drivers of these
340 changes is a daunting task. Studies quantifying global declines are mixed with evidences of more
341 heterogeneous changes, including declines, increases, and no net change at different levels of
342 biological diversity. In a context of accelerated global change, clarifying and addressing these
343 sources of heterogeneity in temporal changes of biodiversity is needed to inform conservation
344 policies. Plus, far from weakening the knowledge on biodiversity loss, working on uncertainty is in
345 fact the best way to consolidate what we know. Here, we explored how different methodological
346 pathways to produce estimates of biodiversity changes likely influence the direction of trends in
347 population abundance, species richness and community composition as well as the effect of the
348 drivers of these trends. As our analysis is based on a single bibliographic database and additions
349 from the grey literature, we recognize that the coverage of the literature is certainly not complete.
350 Moreover, with the explosion of open databases worldwide, the literature on this subject is
351 flourishing and articles are accumulating very quickly (Appendix S2). There is therefore a gap
352 between the representativity of the current knowledge compared to the one from the established
353 corpus. However, the robustness of our methodology to target a broad range of papers within the
354 initial query allows a certain degree of confidence regarding the approximations we may draw in
355 the present paper.

356

357 ***Bottom-up approaches amplify publication bias***

358 Producing biodiversity syntheses requires gathering empirical evidence of biodiversity changes
359 distributed on a broad scale. We have identified two major ways of estimating global changes in
360 biodiversity: either by synthesizing already available information (“bottom-up”) or by producing
361 estimates from raw global data (“top-down”). Our findings show that the bottom-up syntheses from
362 reviews or meta-analyses most often conclude a decline in biodiversity. This could be due to biases
363 in the selected studies when performing such bottom-up assessment. The political intent of
364 governments or conservationists to monitor more endangered species results in a selection bias if
365 the trend of those species is interpreted as an average trend for the entire taxonomic group
366 considered (Boakes et al., 2010). Such selection bias might be further amplified due to a publication

367 bias: as studies generally hypothesize a decline, any study showing neutral or positive change
368 would fail to prove the hypothesis, which may encourage authors to select more results on declining
369 trends (Haddaway et al., 2020; Mlinarić et al., 2017). Studies showing biodiversity declines are thus
370 over-represented compared to studies showing neutral or positive changes. This tendency is
371 exacerbated by bottom-up assessments as they rely on already published materials. We do not imply
372 that reviews or meta-analyses should not be conducted, but rather that biodiversity syntheses should
373 make sure publication bias is taken into account (see Haddaway et al., 2020 for recommendations).

374

375 ***Biodiversity syntheses are challenged by data characteristics***

376 Quantification of trends is also based on empirical biodiversity data that are biased in terms of
377 spatial coverage and taxa considered. Plus, these data often represent monitoring of populations or
378 communities through short sampling periods, hence the challenges around the use and collection of
379 these data. Our findings confirm the major geographical and taxonomic drawbacks when estimates
380 are generated from empirical data, namely geographical and taxonomic ones. Europe and North
381 America are the most assessed continents (Boakes et al., 2010; Manes et al., 2021; Saha et al.,
382 2018), whereas information is least available in the tropics (Feeley & Silman, 2011; Saha et al.,
383 2018) and in regions that are currently under pressure (Pereira et al., 2012). Regarding taxonomic
384 groups, terrestrial organisms, and especially vertebrates (Davison et al., 2021; Pereira et al., 2012;
385 Theobald et al., 2015), are more studied than marine ones (Manes et al., 2021). Overall, there is an
386 over-representation of endangered species (Boakes et al., 2010; Saha et al., 2018). In addition to
387 these already well-documented taxonomic and geographical biases, we show that the length of the
388 period considered is influencing the conclusions: the reviewed studies that are based on short time
389 series identify more declines. If species historically monitored are recovering, short time series may
390 miss their recovery. Selecting longer time-series should buffer the decline in this scenario. These
391 results contradict other findings though. For instance, Vellend et al. (2013) found that the length of
392 the time series had no effect on the assessment of biodiversity change. On the contrary, Gonzalez et
393 al. (2016) showed that the percentage of decline increased with the length of the time series.

394

395 Thus, the heterogeneity in the conclusions is probably related to heterogeneity in the data
396 characteristics. Beyond being a simple source of heterogeneity, this indicates a lack of
397 representativeness that may represent a bias and therefore influence the reliability of the estimated
398 trends. The lack of representativity in certain groups or regions does not ensure the reliability of the
399 trends, but recent studies highlighted that accounting for these biases (e.g. through weighting
400 processes (McRae et al., 2017)) is not sufficient to correctly assess the trends (Dove et al., 2023).

401 They showed that not only short time series are less reliable than longer ones (Wauchope et al.,
402 2019), but also that the assessment of temporal changes in biodiversity at the global scale depends
403 more on the number of time series (considered here at population scale) than on the
404 representativeness of the number of species present within each taxonomic group in the data (Dove
405 et al., 2023).

406

407 To overcome these issues, a short-term solution would be to merge existing aggregated databases
408 that are complimentary in order to increase the amount of data; although it should be noted that
409 such synthesis requires caution regarding possible scale mismatches among datasets (temporal
410 and/or spatial) and diversity in the metrics used (Record et al., 2021). Additionally, sensitivity
411 analyses regarding the length of the time series used should be implemented systematically. The
412 long-term solution is obviously to invest in maintaining monitoring schemes for collecting data in
413 the long term and to make a special effort on covering overlooked taxa and areas.

414

415

416 ***Beyond linear trends***

417 Another part of the observed heterogeneities arise from methodological issues. In particular, we
418 identify two main methodological approaches. The first is the use of global indicators, and most
419 notably in the papers reviewed here the use of the LPI as well as the RLI. The second approach is
420 the use of individual models, very often linear, in order to characterize trends in time series
421 evaluating population abundances or species richness.

422

423 We highlight the fact that the ecological level considered is not as important in the heterogeneities
424 as might be thought, but that the aggregation of trends into single metrics masks the heterogeneities,
425 providing an abundant proportion of declining results. Such indicators have already been criticized
426 for this reason, but also because of their sensitivity to random fluctuations and data gaps (Buschke
427 et al., 2021; Leung et al., 2020).

428

429 Our results show that linear models applied to individual time series seem to be a less biased
430 modelling strategy compared to global indicators. Linear models are often used to estimate the rate
431 of change of a variable (Christensen et al., 2014; Donald et al., 2001; Sánchez-Bayo & Wyckhuys,
432 2019), the most widely used variable being species richness (Hillebrand et al., 2018). However,
433 species richness may not be the most appropriate variable to measure biodiversity change. Indeed,
434 beyond being widely used, changes in species richness remain poorly informative. It has been

435 proven to be insensitive to other form of biodiversity change (Hillebrand et al., 2018; Santini et al.,
436 2017) and unreliable to detect direction in trends despite the relative simplicity to calculate it
437 (Valdez et al., 2023).

438

439 Plus, focusing on linear trends may hide other relevant components. First, there are many examples
440 of non-linear dynamics in nature, both in pressures (Steffen et al., 2015) and in responses (McGill et
441 al., 2015). Linear trends are also more likely to miss different periods within time series (e.g. the
442 recent recovery of a given species monitored for a long time). For these reasons, describing non-
443 linear patterns should be even more straightforward than using simplistic linear approaches. Some
444 complex methods like generalised linear models with polynomial regression splines (Cunningham
445 & Olsen, 2009) or generalised additive models (GAM) (Buckland et al., 2005; Fewster et al., 2000)
446 are already being used to describe non-linear dynamics. However, we suggest here to use simpler
447 workflows, like the one described by Rigal et al. (2020) for instance, to avoid overfitting and allow
448 comparisons between different species.

449

450 Also, most attention is given on trends while other characteristics in the pattern of biodiversity
451 changes are most often ignored. Variability (and changes thereof) is a proxy of stability in
452 ecological systems (Donohue et al., 2016). Yet, the variability in biodiversity dynamics is largely
453 overlooked: variability is often studied at the scale of ecosystems or communities to characterize
454 their stability (Hughes et al., 2013; Scheffer et al., 2009), but very little at the scale of populations
455 in the context of a global analysis. In the studies we reviewed, stability has only been investigated
456 once (Marsh, 2001), and the few examples of studies that have assessed population stability at the
457 global scale involve taxonomically and spatially biased data (Leung et al., 2017; Williams et al.,
458 2022). Instead, most studies focus on the extinction of species (Bellard et al., 2012) and ignore
459 fluctuations in populations, although fluctuations can indicate high vulnerability to extinction
460 (Clements et al., 2015).

461

462 ***Integrating interacting drivers into biodiversity monitoring and changes assessments***

463 Understanding the biodiversity crisis also requires to understand how global change drivers impact
464 biodiversity across spatial and temporal scales. There are many local or regional examples in which
465 pressure-response links have been established, especially with habitat destruction or land-use
466 practices (Donald et al., 2001; Kohsaka et al., 2013; Liu et al., 2013; Nowakowski et al., 2017), but
467 our work reveals that the way in which pressures aggregate on a global scale and impact the spatio-
468 temporal dynamics of biodiversity remains poorly understood. We also highlight that while links to

469 climate change can be established (Comte & Lenoir, 2020; Knape & de Valpine, 2012; Parmesan &
470 Yohe, 2003), the difficulty of making links between other anthropogenic pressures *and* biodiversity
471 changes at the same time remains (only few examples, e.g. Mantyka-pringle et al., 2012; Nunez &
472 Alkemade, 2021; Oliver & Morecroft, 2014). This observation confirms Mazor et al.(2018): they
473 found that 40,3% of the research effort on drivers of biodiversity loss focus on climate change,
474 while only 5,4% focus on pollution and 5% on overexploitation. However, many studies have
475 suggested that climate change and other threats to biodiversity may interact to lead to even greater
476 consequences (Bowler et al., 2020; Brook et al., 2008; Oliver & Morecroft, 2014; Sala et al., 2000).
477 Such lack of global scale integration jeopardises our understanding of the human induced drivers of
478 biodiversity changes at large scale, leading to inappropriate management strategies and missed
479 conservation opportunities (Sirami et al., 2017).

480

481 ***Conservation perspectives***

482 Although some of the biases we report here potentially lead to over or under estimate the overall
483 decline in biodiversity, we do not question the magnitude of the biodiversity crisis. We are now
484 moving forward with the post-2020 Global Biodiversity Framework and all the Aichi Targets for
485 2020 have been only partially achieved or not achieved at all. The Intergovernmental Platform on
486 Biodiversity and Ecosystem Services (IPBES) has published a very alarming global biodiversity
487 assessment report in 2019. In this context, characterizing the state of biodiversity, the impact of
488 drivers and responses is a key step to take action and “bend the curve of biodiversity loss” (Tekwa
489 et al., 2023). While we urgently need reliable assessments to quantify temporal changes in
490 biodiversity and their links to global drivers, we call for more attention to overlooked and yet
491 informative components of biodiversity changes.

492

493 For instance, the Group on Earth Observations Biodiversity Observation Network (GEO BON)
494 initiative emerged in 2013 with the concept of “Essential Biodiversity Variables” defined as
495 “measurement required for study, reporting, and management of biodiversity change”, focusing on
496 the status and trend of biodiversity components (Pereira et al., 2013). These metrics transform data
497 from a variety of sources into indicators that provide a synthetic description of different levels of
498 biodiversity organization, thus facilitating the translation of biodiversity data into policy
499 information.

500

501 We suggest that conservation actions should be based not only on Essential Biodiversity Variables
502 but more globally on Essential Biodiversity Data, with requirements on taxonomic, geographic and

503 temporal coverage ensuring the reliability of estimated trends and thus being able to guide strategies
504 based on the least biased observations possible. Similarly, the implementation of a framework such
505 as Essential Biodiversity Assessment Methods, including the need to systematically take into
506 account different levels of biodiversity, as well as the measurement not only of linear dynamics,
507 should be considered collectively and on a large scale.

508

509 Eventually, examining the drivers of temporal changes of biodiversity also provides evidence for
510 conservation decision-making (Ehrlén & Morris, 2015; Hefley et al., 2016). Conservation can only
511 be considered in conjunction with an examination of the drivers of these trends, which are of course
512 complex and heterogeneous. Our results plead for an urgent need to develop guidance on the
513 necessary quality required for drivers data, their spatial and temporal coverage, the number of
514 drivers to be considered, and their identity. Methods establishing links also need to be considered,
515 and perspectives are to be explored in terms of not only correlative but also causal links (Rigal et
516 al., 2023).

517

518 The results of our review confirm the idea that a multifaceted view of biodiversity is needed to
519 capture all trajectories and the risk of relying solely on global indicators. Empirical evidence for the
520 ongoing biodiversity crisis will never reduce to a silver bullet and univocal metric of global
521 biodiversity change. Eventually, denialism and inaction can be encouraged by the fiction that the
522 state and fate of global biodiversity can be encapsulated in a given metric. A pernicious effect of
523 relying on global metrics would be to consider that the situation is satisfactory providing that a
524 given metric is stable if declining populations are “compensated” or “balanced” by increasing ones.
525 But declines and increases in specific components of biodiversity caused by human activities are by
526 no means cancelling out each other. Any decline of a population or an extinction of a species caused
527 by human activities is a conservation and ethical concern. By quantifying the nuance and full
528 distribution of the impacts of drivers on temporal changes of biodiversity, we should better
529 understand ongoing changes in biodiversity and make sure that conservation actions are making
530 differences.

531

532 **Data, scripts, code, and supplementary information availability**

533 The R code for metadata manipulation and visualization is available on GitHub
534 (<https://github.com/MaelysBoennec/Sources-of-confusion-in-global-biodiversity-trends>).

535 **Conflict of interest disclosure**

536 The authors declare that they comply with the PCI rule of having no financial conflicts of interest in
537 relation to the content of the article.

538

539 **References**

Agardy, T. (2005). Global marine conservation policy versus site-level implementation : The mismatch of scale and its implications. *Marine Ecology-progress Series - MAR ECOLOGICAL PROGRESS SER*, 300, 242-248. <https://doi.org/10.3354/meps300242>

Bellard, C., Bertelsmeier, C., Leadley, P., Thuiller, W., & Courchamp, F. (2012). Impacts of climate change on the future of biodiversity. *Ecology Letters*, 15(4), 365-377. <https://doi.org/10.1111/j.1461-0248.2011.01736.x>

Blowes, S. A., Supp, S. R., Antão, L. H., Bates, A., Bruelheide, H., Chase, J. M., Moyes, F., Magurran, A., McGill, B., Myers-Smith, I. H., Winter, M., Bjorkman, A. D., Bowler, D. E., Byrnes, J. E. K., Gonzalez, A., Hines, J., Isbell, F., Jones, H. P., Navarro, L. M., ... Dornelas, M. (2019). The geography of biodiversity change in marine and terrestrial assemblages. *Science*, 366(6463), 339-345. <https://doi.org/10.1126/science.aaw1620>

Boakes, E. H., McGowan, P. J. K., Fuller, R. A., Chang-qing, D., Clark, N. E., O'Connor, K., & Mace, G. M. (2010). Distorted Views of Biodiversity : Spatial and Temporal Bias in Species Occurrence Data. *PLOS Biology*, 8(6), e1000385. <https://doi.org/10.1371/journal.pbio.1000385>

Bowler, D. E., Bjorkman, A. D., Dornelas, M., Myers-Smith, I. H., Navarro, L. M., Niamir, A., Supp, S. R., Waldock, C., Winter, M., Vellend, M., Blowes, S. A., Böhning-Gaese, K., Bruelheide, H., Elahi, R., Antão, L. H., Hines, J., Isbell, F., Jones, H. P., Magurran, A. E., ... Bates, A. E. (2020). Mapping human pressures on biodiversity across the planet uncovers anthropogenic threat complexes. *People and Nature*, 2(2), 380-394. <https://doi.org/10.1002/pan3.10071>

Brook, B., Sodhi, N., & Bradshaw, C. (2008). Synergies among extinction drivers under global change. *Trends in ecology & evolution*, 23, 453-460. <https://doi.org/10.1016/j.tree.2008.03.011>

Brooke, M. de L., Butchart, S. H. M., Garnett, S. T., Crowley, G. M., Mantilla-Beniers, N. B., & Stattersfield, A. J. (2008). Rates of movement of threatened bird species between IUCN red list categories and toward extinction. *Conservation Biology: The Journal of the Society for Conservation Biology*, 22(2), 417-427. <https://doi.org/10.1111/j.1523-1739.2008.00905.x>

- Buckland, S. t, Magurran, A. e, Green, R. e, & Fewster, R. m. (2005). Monitoring change in biodiversity through composite indices. *Philosophical Transactions of the Royal Society B: Biological Sciences*, 360(1454), 243-254. <https://doi.org/10.1098/rstb.2004.1589>
- Buschke, F., Hagan, J., Santini, L., & Coetzee, B. (2021). Random population fluctuations bias the Living Planet Index. *Nature Ecology & Evolution*, 5. <https://doi.org/10.1038/s41559-021-01494-0>
- Capdevila, P., Noviello, N., McRae, L., Freeman, R., & Clements, C. F. (2022). Global patterns of resilience decline in vertebrate populations. *Ecology Letters*, 25(1), 240-251. <https://doi.org/10.1111/ele.13927>
- Cardinale, B. (2014). Overlooked local biodiversity loss. *Science*, 344(6188), 1098-1098. <https://doi.org/10.1126/science.344.6188.1098-a>
- Cardinale, B. J., Gonzalez, A., Allington, G. R. H., & Loreau, M. (2018). Is local biodiversity declining or not? A summary of the debate over analysis of species richness time trends. *Biological Conservation*, 219, 175-183. <https://doi.org/10.1016/j.biocon.2017.12.021>
- Ceballos, G., Ehrlich, P. R., Barnosky, A. D., García, A., Pringle, R. M., & Palmer, T. M. (2015). Accelerated modern human-induced species losses : Entering the sixth mass extinction. *Science Advances*, 1(5), e1400253. <https://doi.org/10.1126/sciadv.1400253>
- Christensen, V., Coll, M., Piroddi, C., Steenbeek, J., Buszowski, J., & Pauly, D. (2014). A century of fish biomass decline in the ocean. *Marine Ecology Progress Series*, 512. <https://doi.org/10.3354/meps10946>
- Clements, C. F., Drake, J. M., Griffiths, J. I., & Ozgul, A. (2015). Factors Influencing the Detectability of Early Warning Signals of Population Collapse. *The American Naturalist*, 186(1), 50-58. <https://doi.org/10.1086/681573>
- Comte, L., & Lenoir, J. (2020). Decoupled land–sea biodiversity trends. *Nature Ecology & Evolution*, 4(7), Article 7. <https://doi.org/10.1038/s41559-020-1191-9>
- Cunningham, R., & Olsen, P. (2009). A statistical methodology for tracking long-term change in reporting rates of birds from volunteer-collected presence–absence data. *Biodiversity and Conservation*, 18(5), 1305-1327. <https://doi.org/10.1007/s10531-008-9509-y>
- Daskalova, G. N., Myers-Smith, I. H., Bjorkman, A. D., Blowes, S. A., Supp, S. R., Magurran, A. E., & Dornelas, M. (2020). Landscape-scale forest loss as a catalyst of population and biodiversity change. *Science*, 368(6497), 1341-1347. <https://doi.org/10.1126/science.aba1289>

- Davison, C. W., Rahbek, C., & Morueta-Holme, N. (2021). Land-use change and biodiversity : Challenges for assembling evidence on the greatest threat to nature. *Global Change Biology*, 27(21), 5414-5429. <https://doi.org/10.1111/gcb.15846>
- Donald, P. F., Green, R., & Heath, M. F. (2001). Agricultural intensification and the collapse of Europe's farmland bird populations. *Proceedings of the Royal Society, Series B*, 155, 39-43.
- Donohue, I., Hillebrand, H., Montoya, J. M., Petchey, O. L., Pimm, S. L., Fowler, M. S., Healy, K., Jackson, A. L., Lurgi, M., McClean, D., O'Connor, N. E., O'Gorman, E. J., & Yang, Q. (2016). Navigating the complexity of ecological stability. *Ecology Letters*, 19(9), 1172-1185. <https://doi.org/10.1111/ele.12648>
- Dornelas, M., Antão, L. H., Moyes, F., Bates, A. E., Magurran, A. E., Adam, D., Akhmetzhanova, A. A., Appeltans, W., Arcos, J. M., Arnold, H., Ayyappan, N., Badihi, G., Baird, A. H., Barbosa, M., Barreto, T. E., Bässler, C., Bellgrove, A., Belmaker, J., Benedetti-Cecchi, L., ... Zettler, M. L. (2018). BioTIME : A database of biodiversity time series for the Anthropocene. *Global Ecology and Biogeography*, 27(7), 760-786. <https://doi.org/10.1111/geb.12729>
- Dornelas, M., Gotelli, N. J., McGill, B., Shimadzu, H., Moyes, F., Sievers, C., & Magurran, A. E. (2014). Assemblage Time Series Reveal Biodiversity Change but Not Systematic Loss. *Science*, 344(6181), 296-299. <https://doi.org/10.1126/science.1248484>
- Dornelas, M., Gotelli, N. J., Shimadzu, H., Moyes, F., Magurran, A. E., & McGill, B. J. (2019). A balance of winners and losers in the Anthropocene. *Ecology Letters*, 22(5), 847-854. <https://doi.org/10.1111/ele.13242>
- Dove, S., Böhm, M., Freeman, R., McRae, L., & Murrell, D. J. (2023). *How much data do we need? Reliability and data deficiency in global vertebrate biodiversity trends* (p. 2023.03.18.532273). bioRxiv. <https://doi.org/10.1101/2023.03.18.532273>
- Du, Y., Yang, B., Chen, S.-C., & Ma, K. (2019). Diverging shifts in spring phenology in response to biodiversity loss in a subtropical forest. *Journal of Vegetation Science*, 30(6), 1175-1183. <https://doi.org/10.1111/jvs.12806>
- Duchenne, F., Porcher, E., Mihoub, J.-B., Lois, G., & Fontaine, C. (2022). Controversy over the decline of arthropods : A matter of temporal baseline? *Peer Community Journal*, 2. <https://doi.org/10.24072/pcjournal.131>
- Ehrlén, J., & Morris, W. F. (2015). Predicting changes in the distribution and abundance of species under environmental change. *Ecology Letters*, 18(3), 303-314. <https://doi.org/10.1111/ele.12410>

- Erauskin-Extramiana, M., Arrizabalaga, H., Cabré, A., Coelho, R., Rosa, D., Ibaibarriaga, L., & Chust, G. (2020). Are shifts in species distribution triggered by climate change? A swordfish case study. *Deep Sea Research Part II: Topical Studies in Oceanography*, 175, 104666. <https://doi.org/10.1016/j.dsr2.2019.104666>
- Fahrig, L. (2017). Forty years of bias in habitat fragmentation research. In *Effective Conservation Science*. Oxford University Press. <https://doi.org/10.1093/oso/9780198808978.003.0005>
- Feeley, K. J., & Silman, M. R. (2011). The data void in modeling current and future distributions of tropical species. *Global Change Biology*, 17(1), 626-630. <https://doi.org/10.1111/j.1365-2486.2010.02239.x>
- Fewster, R. M., Buckland, S. T., Siriwardena, G. M., Baillie, S. R., & Wilson, J. D. (2000). Analysis of Population Trends for Farmland Birds Using Generalized Additive Models. *Ecology*, 81(7), 1970-1984. <https://doi.org/10.2307/177286>
- Geldmann, J., Manica, A., Burgess, N. D., Coad, L., & Balmford, A. (2019). A global-level assessment of the effectiveness of protected areas at resisting anthropogenic pressures. *Proceedings of the National Academy of Sciences*, 116(46), 23209-23215. <https://doi.org/10.1073/pnas.1908221116>
- Gilman, S. E., Urban, M. C., Tewksbury, J., Gilchrist, G. W., & Holt, R. D. (2010). A framework for community interactions under climate change. *Trends in Ecology & Evolution*, 25(6), 325-331. <https://doi.org/10.1016/j.tree.2010.03.002>
- Gonzalez, A., Cardinale, B. J., Allington, G. R. H., Byrnes, J., Arthur Endsley, K., Brown, D. G., Hooper, D. U., Isbell, F., O'Connor, M. I., & Loreau, M. (2016). Estimating local biodiversity change : A critique of papers claiming no net loss of local diversity. *Ecology*, 97(8), 1949-1960. <https://doi.org/10.1890/15-1759.1>
- Haddad, N. M., Gonzalez, A., Brudvig, L. A., Burt, M. A., Levey, D. J., & Damschen, E. I. (2017). Experimental evidence does not support the Habitat Amount Hypothesis. *Ecography*, 40(1), 48-55. <https://doi.org/10.1111/ecog.02535>
- Haddaway, N. R., Bethel, A., Dicks, L. V., Koricheva, J., Macura, B., Petrokofsky, G., Pullin, A. S., Savilaakso, S., & Stewart, G. B. (2020). Eight problems with literature reviews and how to fix them. *Nature Ecology & Evolution*, 4(12), Article 12. <https://doi.org/10.1038/s41559-020-01295-x>
- He, F., Zarfl, C., Bremerich, V., David, J. N. W., Hogan, Z., Kalinkat, G., Tockner, K., & Jähnig, S. C. (2019). The global decline of freshwater megafauna. *Global Change Biology*, 25(11), 3883-3892. <https://doi.org/10.1111/gcb.14753>

- Hefley, T. J., Hooten, M. B., Drake, J. M., Russell, R. E., & Walsh, D. P. (2016). When can the cause of a population decline be determined? *Ecology Letters*, 19(11), 1353-1362.
<https://doi.org/10.1111/ele.12671>
- Hillebrand, H., Blasius, B., Borer, E. T., Chase, J. M., Downing, J. A., Eriksson, B. K., Filstrup, C. T., Harpole, W. S., Hodapp, D., Larsen, S., Lewandowska, A. M., Seabloom, E. W., Van de Waal, D. B., & Ryabov, A. B. (2018). Biodiversity change is uncoupled from species richness trends : Consequences for conservation and monitoring. *Journal of Applied Ecology*, 55(1), 169-184. <https://doi.org/10.1111/1365-2664.12959>
- Houlahan, J. E., Findlay, C. S., Schmidt, B. R., Meyer, A. H., & Kuzmin, S. L. (2000). Quantitative evidence for global amphibian population declines. *Nature*, 404(6779), 752-755.
<https://doi.org/10.1038/35008052>
- Howe, L. C., MacInnis, B., Krosnick, J. A., Markowitz, E. M., & Socolow, R. (2019). Acknowledging uncertainty impacts public acceptance of climate scientists' predictions. *Nature Climate Change*, 9(11), Article 11. <https://doi.org/10.1038/s41558-019-0587-5>
- Hughes, T. P., Carpenter, S., Rockström, J., Scheffer, M., & Walker, B. (2013). Multiscale regime shifts and planetary boundaries. *Trends in Ecology & Evolution*, 28(7), 389-395.
<https://doi.org/10.1016/j.tree.2013.05.019>
- Inchausti, P., & Halley, J. (2001). Investigating Long-Term Ecological Variability Using the Global Population Dynamics Database. *Science*, 293(5530), 655-657.
<https://doi.org/10.1126/science.293.5530.655>
- IPBES. (2019). *Summary for policymakers of the global assessment report on biodiversity and ecosystem services of the Intergovernmental Science-Policy Platform on Biodiversity and Ecosystem Services*. <https://doi.org/10.5281/zenodo.3553458>
- IPCC. (2021). Summary for Policymakers. In : *Climate Change 2021 : The Physical Science Basis. Contribution of Working Group I to the Sixth Assessment Report of the Intergovernmental Panel on Climate Change* [Masson-Delmotte, V., P. Zhai, A. Pirani, S.L. Connors, C. Péan, S. Berger, N. Caud, Y. Chen, L. Goldfarb, M.I. Gomis, M. Huang, K. Leitzell, E. Lonnoy, J.B.R. Matthews, T.K. Maycock, T. Waterfield, O. Yelekçi, R. Yu, and B. Zhou (eds.)]. Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA, pp. 3–32,. doi:10.1017/9781009157896.001.
- IUCN 2022. *The IUCN Red List of Threatened Species. Version 2021-3*. (s. d.). IUCN Red List of Threatened Species. Consulté 16 mai 2022, à l'adresse <https://www.iucnredlist.org>
- Joslyn, S. L., & LeClerc, J. E. (2016). Climate Projections and Uncertainty Communication. *Topics in Cognitive Science*, 8(1), 222-241. <https://doi.org/10.1111/tops.12177>

- Keith, D., Akçakaya, H. R., Butchart, S. H. M., Collen, B., Dulvy, N. K., Holmes, E. E., Hutchings, J. A., Keinath, D., Schwartz, M. K., Shelton, A. O., & Waples, R. S. (2015). Temporal correlations in population trends : Conservation implications from time-series analysis of diverse animal taxa. *Biological Conservation*, *192*, 247-257.
<https://doi.org/10.1016/j.biocon.2015.09.021>
- Knape, J., & de Valpine, P. (2012). Are patterns of density dependence in the Global Population Dynamics Database driven by uncertainty about population abundance? *Ecology Letters*, *15*(1), 17-23. <https://doi.org/10.1111/j.1461-0248.2011.01702.x>
- Kohsaka, R., Shih, W., Saito, O., & Sadohara, S. (2013). Local Assessment of Tokyo : Satoyama and Satoumi – Traditional Landscapes and Management Practices in a Contemporary Urban Environment. In T. Elmqvist, M. Fragkias, J. Goodness, B. Güneralp, P. J. Marcotullio, R. I. McDonald, S. Parnell, M. Schewenius, M. Sendstad, K. C. Seto, & C. Wilkinson (Éds.), *Urbanization, Biodiversity and Ecosystem Services : Challenges and Opportunities : A Global Assessment* (p. 93-105). Springer Netherlands. https://doi.org/10.1007/978-94-007-7088-1_8
- Koleček, J., Reif, J., Šálek, M., Hanzelka, J., Sottas, C., & Kubelka, V. (2021). Global population trends in shorebirds : Migratory behaviour makes species at risk. *The Science of Nature*, *108*.
<https://doi.org/10.1007/s00114-021-01717-1>
- Laurance, W. F. (2007). Have we overstated the tropical biodiversity crisis? *Trends in Ecology & Evolution*, *22*(2), 65-70. <https://doi.org/10.1016/j.tree.2006.09.014>
- Leiserowitz, A., Maibach, E. W., Roser-Renouf, C., Feinberg, G., & Howe, P. (2013). *Climate Change in the American Mind : Americans' Global Warming Beliefs and Attitudes in April 2013* (SSRN Scholarly Paper N° 2298705). <https://doi.org/10.2139/ssrn.2298705>
- Leung, B., Greenberg, D. A., & Green, D. M. (2017). Trends in mean growth and stability in temperate vertebrate populations. *Diversity and Distributions*, *23*(12), 1372-1380.
<https://doi.org/10.1111/ddi.12636>
- Leung, B., Hargreaves, A. L., Greenberg, D. A., McGill, B., Dornelas, M., & Freeman, R. (2020). Clustered versus catastrophic global vertebrate declines. *Nature*, *588*(7837), 267-271.
<https://doi.org/10.1038/s41586-020-2920-6>
- Liu, W., Chen, X., & Wang, Q. (2013). Local Assessment of Shanghai : Effects of Urbanization on the Diversity of Macrobenthic Invertebrates. In T. Elmqvist, M. Fragkias, J. Goodness, B. Güneralp, P. J. Marcotullio, R. I. McDonald, S. Parnell, M. Schewenius, M. Sendstad, K. C. Seto, & C. Wilkinson (Éds.), *Urbanization, Biodiversity and Ecosystem Services :*

- Challenges and Opportunities : A Global Assessment* (p. 107-122). Springer Netherlands.
https://doi.org/10.1007/978-94-007-7088-1_9
- Loh, J., Green, R., Ricketts, T., Lamoreux, J., Jenkins, M., Kapos, V., & Randers, J. (2005). The Living Planet Index : Using species population time series to track trends in biodiversity. *Philosophical transactions of the Royal Society of London. Series B, Biological sciences*, 360, 289-295. <https://doi.org/10.1098/rstb.2004.1584>
- Manes, S., Costello, M. J., Beckett, H., Debnath, A., Devenish-Nelson, E., Grey, K.-A., Jenkins, R., Khan, T. M., Kiessling, W., Krause, C., Maharaj, S. S., Midgley, G. F., Price, J., Talukdar, G., & Vale, M. M. (2021). Endemism increases species' climate change risk in areas of global biodiversity importance. *Biological Conservation*, 257, 109070.
<https://doi.org/10.1016/j.biocon.2021.109070>
- Mantyka-pringle, C. S., Martin, T. G., & Rhodes, J. R. (2012). Interactions between climate and habitat loss effects on biodiversity : A systematic review and meta-analysis. *Global Change Biology*, 18(4), 1239-1252. <https://doi.org/10.1111/j.1365-2486.2011.02593.x>
- Marsh, D. M. (2001). Fluctuations in amphibian populations : A meta-analysis. *Biological Conservation*, 101(3), 327-335. [https://doi.org/10.1016/S0006-3207\(01\)00076-3](https://doi.org/10.1016/S0006-3207(01)00076-3)
- Mazor, T., Doropoulos, C., Schwarzmüller, F., Gladish, D. W., Kumaran, N., Merkel, K., Di Marco, M., & Gagic, V. (2018). Global mismatch of policy and research on drivers of biodiversity loss. *Nature Ecology & Evolution*, 2(7), Article 7.
<https://doi.org/10.1038/s41559-018-0563-x>
- McGill, B. J., Dornelas, M., Gotelli, N. J., & Magurran, A. E. (2015). Fifteen forms of biodiversity trend in the Anthropocene. *Trends in Ecology & Evolution*, 30(2), 104-113.
<https://doi.org/10.1016/j.tree.2014.11.006>
- Mckinney, M., & Lockwood, J. (1999). Biotic Homogenization : A Few Winners Replacing Many Losers in the Next Mass Extinction. *Trends in Ecology & Evolution*, 14, 450-453.
[https://doi.org/10.1016/S0169-5347\(99\)01679-1](https://doi.org/10.1016/S0169-5347(99)01679-1)
- McRae, L., Deinet, S., & Freeman, R. (2017). The Diversity-Weighted Living Planet Index : Controlling for Taxonomic Bias in a Global Biodiversity Indicator. *PLOS ONE*, 12(1), e0169156. <https://doi.org/10.1371/journal.pone.0169156>
- Mihoub, J.-B., Henle, K., Titeux, N., Brotons, L., Brummitt, N. A., & Schmeller, D. S. (2017). Setting temporal baselines for biodiversity : The limits of available monitoring data for capturing the full impact of anthropogenic pressures. *Scientific Reports*, 7(1), Article 1.
<https://doi.org/10.1038/srep41591>

- Mlinarić, A., Mlinarić, A., Horvat, M., Horvat, M., Smolčić, V. Š., Smolčić, V. Š., & Smolčić, V. Š. (2017). Dealing with the positive publication bias : Why you should really publish your negative results. *Biochemia Medica*, 27(3), 0-0. <https://doi.org/10.11613/BM.2017.030201>
- Mora, C., & Sale, P. F. (2011). Ongoing global biodiversity loss and the need to move beyond protected areas : A review of the technical and practical shortcomings of protected areas on land and sea. *Marine Ecology Progress Series*, 434, 251-266. <https://doi.org/10.3354/meps09214>
- Nowakowski, A. J., Thompson, M. E., Donnelly, M. A., & Todd, B. D. (2017). Amphibian sensitivity to habitat modification is associated with population trends and species traits. *Global Ecology and Biogeography*, 26(6), 700-712. <https://doi.org/10.1111/geb.12571>
- Nunez, S., & Alkemade, R. (2021). Exploring interaction effects from mechanisms between climate and land-use changes and the projected consequences on biodiversity. *Biodiversity and Conservation*, 30(12), 3685-3696. <https://doi.org/10.1007/s10531-021-02271-y>
- Olden, J. D., Leroy Poff, N., Douglas, M. R., Douglas, M. E., & Fausch, K. D. (2004). Ecological and evolutionary consequences of biotic homogenization. *Trends in Ecology & Evolution*, 19(1), 18-24. <https://doi.org/10.1016/j.tree.2003.09.010>
- Oliver, T. H., & Morecroft, M. D. (2014). Interactions between climate change and land use change on biodiversity : Attribution problems, risks, and opportunities. *WIREs Climate Change*, 5(3), 317-335. <https://doi.org/10.1002/wcc.271>
- Paprocki, N., Heath, J. A., & Novak, S. J. (2014). Regional distribution shifts help explain local changes in wintering raptor abundance : Implications for interpreting population trends. *PloS One*, 9(1), e86814. <https://doi.org/10.1371/journal.pone.0086814>
- Parmesan, C., & Yohe, G. (2003). A globally coherent fingerprint of climate change impacts across natural systems. *Nature*, 421(6918), Article 6918. <https://doi.org/10.1038/nature01286>
- Pereira, H. M., Ferrier, S., Walters, M., Geller, G. N., Jongman, R. H. G., Scholes, R. J., Bruford, M. W., Brummitt, N., Butchart, S. H. M., Cardoso, A. C., Coops, N. C., Dulloo, E., Faith, D. P., Freyhof, J., Gregory, R. D., Heip, C., Höft, R., Hurtt, G., Jetz, W., ... Wegmann, M. (2013). Essential Biodiversity Variables. *Science*, 339(6117), 277-278. <https://doi.org/10.1126/science.1229931>
- Pereira, H. M., Navarro, L. M., & Martins, I. S. (2012). Global Biodiversity Change : The Bad, the Good, and the Unknown. *Annual Review of Environment and Resources*, 37(1), 25-50. <https://doi.org/10.1146/annurev-environ-042911-093511>
- Pievani, T. (2014). The sixth mass extinction : Anthropocene and the human impact on biodiversity. *Rendiconti Lincei*, 25(1), 85-93. <https://doi.org/10.1007/s12210-013-0258-9>

- Pimm, S., Raven, P., Peterson, A., Sekercioglu, C. H., & Ehrlich, P. R. (2006). Human impacts on the rates of recent, present, and future bird extinctions. *Proceedings of the National Academy of Sciences of the United States of America*, *103*(29), 10941-10946.
<https://doi.org/10.1073/pnas.06041811103>
- Pörtner, H.-O., Scholes, R. J., Agard, J., Archer, E., Bai, X., Barnes, D., Burrows, M., Chan, L., Cheung, W. L. (William), Diamond, S., Donatti, C., Duarte, C., Eisenhauer, N., Foden, W., Gasalla, M. A., Handa, C., Hickler, T., Hoegh-Guldberg, O., Ichii, K., ... Ngo, H. (2021). *IPBES-IPCC co-sponsored workshop report on biodiversity and climate change*. Zenodo.
<https://doi.org/10.5281/zenodo.5101133>
- Pressey, R. L., Cabeza, M., Watts, M. E., Cowling, R. M., & Wilson, K. A. (2007). Conservation planning in a changing world. *Trends in Ecology & Evolution*, *22*(11), 583-592.
<https://doi.org/10.1016/j.tree.2007.10.001>
- Radchuk, V., Reed, T., Teplitsky, C., van de Pol, M., Charmantier, A., Hassall, C., Adamík, P., Adriaensen, F., Ahola, M. P., Arcese, P., Miguel Avilés, J., Balbontin, J., Berg, K. S., Borrás, A., Burthe, S., Clobert, J., Dehnhard, N., de Lope, F., Dhondt, A. A., ... Kramer-Schadt, S. (2019). Adaptive responses of animals to climate change are most likely insufficient. *Nature Communications*, *10*(1), Article 1. <https://doi.org/10.1038/s41467-019-10924-4>
- Record, S., Voelker, N. M., Zarnetske, P. L., Wisnoski, N. I., Tonkin, J. D., Swan, C., Marazzi, L., Lany, N., Lamy, T., Compagnoni, A., Castorani, M. C. N., Andrade, R., & Sokol, E. R. (2021). Novel Insights to Be Gained From Applying Metacommunity Theory to Long-Term, Spatially Replicated Biodiversity Data. *Frontiers in Ecology and Evolution*, *8*.
<https://www.frontiersin.org/article/10.3389/fevo.2020.612794>
- Reilly, J., Stone, P. H., Forest, C. E., Webster, M. D., Jacoby, H. D., & Prinn, R. G. (2001). Uncertainty and Climate Change Assessments. *Science*.
<https://doi.org/10.1126/science.1062001>
- Rigal, S., Dakos, V., Alonso, H., Auniņš, A., Benkő, Z., Brotons, L., Chodkiewicz, T., Chylarecki, P., de Carli, E., del Moral, J. C., Domşa, C., Escandell, V., Fontaine, B., Foppen, R., Gregory, R., Harris, S., Herrando, S., Husby, M., Ieronymidou, C., ... Devictor, V. (2023). Farmland practices are driving bird population decline across Europe. *Proceedings of the National Academy of Sciences*, *120*(21), e2216573120.
<https://doi.org/10.1073/pnas.2216573120>
- Rigal, S., Devictor, V., & Dakos, V. (2020). A method for classifying and comparing non-linear trajectories of ecological variables. *Ecological Indicators*, *112*, 106113.
<https://doi.org/10.1016/j.ecolind.2020.106113>

- Saha, A., McRae, L., Dodd Jr, C. K., Gadsden, H., Hare, K. M., Lukoschek, V., & Böhm, M. (2018). Tracking Global Population Trends : Population Time-Series Data and a Living Planet Index for Reptiles. *Journal of Herpetology*, 52(3), Article 3.
- Sala, O. E., Stuart Chapin, F., III, Armesto, J. J., Berlow, E., Bloomfield, J., Dirzo, R., Huber-Sanwald, E., Huenneke, L. F., Jackson, R. B., Kinzig, A., Leemans, R., Lodge, D. M., Mooney, H. A., Oesterheld, M., Poff, N. L., Sykes, M. T., Walker, B. H., Walker, M., & Wall, D. H. (2000). Global Biodiversity Scenarios for the Year 2100. *Science*, 287(5459), 1770-1774. <https://doi.org/10.1126/science.287.5459.1770>
- Sánchez-Bayo, F., & Wyckhuys, K. A. G. (2019). Worldwide decline of the entomofauna : A review of its drivers. *Biological Conservation*, 232, 8-27. <https://doi.org/10.1016/j.biocon.2019.01.020>
- Santini, L., Belmaker, J., Costello, M. J., Pereira, H. M., Rossberg, A. G., Schipper, A. M., Ceașu, S., Dornelas, M., Hilbers, J. P., Hortal, J., Huijbregts, M. A. J., Navarro, L. M., Schiffers, K. H., Visconti, P., & Rondinini, C. (2017). Assessing the suitability of diversity metrics to detect biodiversity change. *Biological Conservation*, 213, 341-350. <https://doi.org/10.1016/j.biocon.2016.08.024>
- Sax, D. F., & Gaines, S. D. (2003). Species diversity : From global decreases to local increases. *Trends in Ecology & Evolution*, 18(11), 561-566. [https://doi.org/10.1016/S0169-5347\(03\)00224-6](https://doi.org/10.1016/S0169-5347(03)00224-6)
- Scheffer, M., Bascompte, J., Brock, W., Brovkin, V., Carpenter, S., Dakos, V., Held, H., Nes, E., Rietkerk, M., & Sugihara, G. (2009). Early-Warning Signals for Critical Transitions. *Nature*, 461, 53-59. <https://doi.org/10.1038/nature08227>
- Sirami, C., Caplat, P., Popy, S., Clamens, A., Arlettaz, R., Jiguet, F., Brotons, L., & Martin, J.-L. (2017). Impacts of global change on species distributions : Obstacles and solutions to integrate climate and land use. *Global Ecology and Biogeography*, 26(4), 385-394. <https://doi.org/10.1111/geb.12555>
- Steffen, W., Broadgate, W., Deutsch, L., Gaffney, O., & Ludwig, C. (2015). The trajectory of the Anthropocene : The Great Acceleration. *The Anthropocene Review*, 2(1), 81-98. <https://doi.org/10.1177/2053019614564785>
- Tekwa, E., Gonzalez, A., Zurell, D., & O'Connor, M. (2023). Detecting and attributing the causes of biodiversity change : Needs, gaps and solutions. *Philosophical Transactions of the Royal Society B: Biological Sciences*, 378(1881), 20220181. <https://doi.org/10.1098/rstb.2022.0181>

- Theobald, E. J., Ettinger, A. K., Burgess, H. K., DeBey, L. B., Schmidt, N. R., Froehlich, H. E., Wagner, C., HilleRisLambers, J., Tewksbury, J., Harsch, M. A., & Parrish, J. K. (2015). Global change and local solutions : Tapping the unrealized potential of citizen science for biodiversity research. *Biological Conservation*, *181*, 236-244.
<https://doi.org/10.1016/j.biocon.2014.10.021>
- Valdez, J. W., Callaghan, C. T., Junker, J., Purvis, A., Hill, S. L. L., & Pereira, H. M. (2023). The undetectability of global biodiversity trends using local species richness. *Ecography*, *2023*(3), e06604. <https://doi.org/10.1111/ecog.06604>
- Vellend, M., Baeten, L., Myers-Smith, I. H., Elmendorf, S. C., Beauséjour, R., Brown, C. D., Frenne, P. D., Verheyen, K., & Wipf, S. (2013). Global meta-analysis reveals no net change in local-scale plant biodiversity over time. *Proceedings of the National Academy of Sciences*, *110*(48), 19456-19459. <https://doi.org/10.1073/pnas.1312779110>
- Wagner, D. L., Fox, R., Salcido, D. M., & Dyer, L. A. (2021). A window to the world of global insect declines : Moth biodiversity trends are complex and heterogeneous. *Proceedings of the National Academy of Sciences*, *118*(2), e2002549117.
<https://doi.org/10.1073/pnas.2002549117>
- Walther, G.-R. (2010). Community and ecosystem responses to recent climate change. *Philosophical Transactions of the Royal Society B: Biological Sciences*, *365*(1549), 2019-2024. <https://doi.org/10.1098/rstb.2010.0021>
- Wauchope, H. S., Amano, T., Sutherland, W. J., & Johnston, A. (2019). When can we trust population trends? A method for quantifying the effects of sampling interval and duration. *Methods in Ecology and Evolution*, *10*(12), 2067-2078. <https://doi.org/10.1111/2041-210X.13302>
- Williams, J. J., Freeman, R., Spooner, F., & Newbold, T. (2022). Vertebrate population trends are influenced by interactions between land use, climatic position, habitat loss and climate change. *Global Change Biology*, *28*(3), 797-815. <https://doi.org/10.1111/gcb.15978>
- Willis, K. J., & Bhagwat, S. A. (2009). Biodiversity and Climate Change. *Science*, *326*(5954), 806-807. <https://doi.org/10.1126/science.1178838>
- Wilson, R. J., & Fox, R. (2021). Insect responses to global change offer signposts for biodiversity and conservation. *Ecological Entomology*, *46*(4), 699-717.
<https://doi.org/10.1111/een.12970>
- Wolf, A. A., Zavaleta, E. S., & Selmants, P. C. (2017). Flowering phenology shifts in response to biodiversity loss. *Proceedings of the National Academy of Sciences*, *114*(13), 3463-3468.
<https://doi.org/10.1073/pnas.1608357114>

- WWF. (2020). Living Planet Report 2020—Bending the curve of biodiversity loss. Almond, R.E.A., Grooten M. and Petersen, T. (Eds). WWF, Gland, Switzerland.
- WWF. (2022). Living Planet Report 2022 – Building a nature-positive society. Almond, R.E.A., Grooten, M., Juffe Bignoli, D. & Petersen, T. (Eds). WWF, Gland, Switzerland.