

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

5 (34000)

6 ***Correspondence to:**

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

8

9 **Abstract**

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

27

28



29 **Introduction**

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

45
46 These heterogeneities in results entails several risks if their sources and uncertainties are not
47 addressed. Firstly, it could lead to sub-optimal or, worse, ineffective conservation policies, as many
48 conservation measures rely on the estimation of indicators or discussion of scenarios covering
49 global biodiversity trend for all taxa on a global scale (Agardy, 2005; Pressey et al., 2007).
50 Furthermore, avoiding to question the sources of the heterogeneity of these results and the
51 uncertainties in the conclusions could encourage a "biodiversity-skepticism" by creating the idea
52 that there is a lack of scientific consensus on the existence of a biodiversity crisis. Similarly,
53 climate-skepticism partly emerged from the belief that there were significant disagreements about
54 global warming among scientists (Joslyn & LeClerc, 2016; Leiserowitz et al., 2013). Such distorted
55 perception was reduced by incorporating uncertainties into IPCC reports, clarifying the fact that
56 much of the variability was due to predictive processes rather than fundamental differences in
57 scientific opinion (Howe et al., 2019; Reilly et al., 2001). Thus, conservation science and
58 knowledge on biodiversity loss should also benefit from such clarification and be consolidated by
59 adopting a transparent and quantitative approach to major biases in the global estimates.
60 If these heterogeneous results have also caused a vivid controversy in the scientific community
61 (Cardinale, 2014; Cardinale et al., 2018; Gonzalez et al., 2016), it is to some extent probably
62 resulting from the multiple meanings of the term *biodiversity*. The formal definition largely

63 popularized by the Rio Earth Summit in 1992 equates biodiversity to "the variability among living
64 organisms from all sources [...] this includes diversity within species, between species and of
65 ecosystems" (CBD, 1992). This definition itself creates a lot of confusion (Díaz & Malhi, 2022;
66 Mammola et al., 2023). For example, the WWF considers biodiversity at the population level
67 (WWF, 2020, 2022). In contrast, others (e.g. Dornelas et al., 2014) consider biodiversity at the
68 species or community level. As declines in population sizes and species richness are not necessarily
69 related, both an increase and decrease in "biodiversity" can be concluded depending on the
70 ecological level of interest. Clarifying the trends for each level of biodiversity should, in principle,
71 limit the confusion. In practice however, trends within the same ecological levels show both a
72 decline globally or no net changes; either at species scale (IUCN, 2019; Dornelas et al., 2014;
73 Vellend et al., 2013) or at population scale (Daskalova et al., 2020; He et al., 2019; Leung et al.,
74 2020; Wagner et al., 2021). Beyond the ecological level considered other factors are therefore also
75 generating confusion in the observed results.

76

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

93

94 Acknowledging our current sources of (mis)understanding of the temporal changes of biodiversity
95 and of their drivers at the global scale is urgently needed (Tekwa et al., 2023). International
96 legislation and objectives (such as those discussed at United Nations Biodiversity Conference of the

97 Parties) directly rely on our general knowledge and understanding of global biodiversity dynamics.
98 The objective of this study is therefore to critically examine this general knowledge. More
99 specifically, we want to review how global biodiversity change is currently quantified, and to
100 identify the most salient sources of confusion when assessing biodiversity trends or the effect of its
101 drivers.

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

118 Here, we conduct a literature review to clarify those sources of confusion, and to review the
119 remaining challenges for future research and conservation science.

120

121 **Methods**

122

123 ***Literature review***

124 We conducted a literature review aiming to identify papers providing an assessment of recent
125 temporal changes of biodiversity globally. We were looking both for papers that studied the changes
126 themselves, but also those that sought to explain the changes by studying the impact of the drivers.
127 Different searches were initially launched in the *Web of Science*. We tested different search terms to
128 refine the results. We wanted to minimise the number of irrelevant references while ensuring that
129 some commonly known relevant articles (e.g. Dornelas et al., 2019) fell within our scope. In the
130 process, we identified broad terms (e.g. "population") that encompassed areas of research beyond

131 our topic. We also found that restricting the search to terms in the titles and abstracts, rather than in
132 the topics of the articles in general, allowed us to limit the search to a manageable number of
133 articles (Appendix S1). The final search we launched in the *Web of Science* on the 08/03/2022 under
134 the institution of the University of Montpellier was thus: TI=((biodiversity OR population* OR
135 communit* OR indicator* OR natur* OR richness OR species OR "biological diversity" OR
136 abundance OR assemblage OR *flora OR *fauna) AND (trend* OR dynamic* OR "time series" OR
137 declin* OR loss OR extinct* OR increas* OR gain OR coloni* OR change* OR fluctuat* OR
138 trajector* OR tempo*)) AND AB=((("temporal" OR time) AND ("analys*" OR "model*" OR stud*
139 OR quantifi*)) NOT TI=("human population" OR "urban population") AND (TI=(global OR
140 worldwide) OR AB=(global OR worldwide)) NOT WC=("meteorology atmospheric sciences" OR
141 "infectious disease" OR "biochemistry molecular biology" OR "paleontology" OR "microbiology").
142 TI is title, AB is abstract and WC is Web of Science Categories. This query resulted in 2,008
143 matches.

144

145 Each of these papers was then reviewed individually by titles, abstracts and then subjected to a full
146 review to check their relevance to our study objective. We included studies that either assess or
147 discuss the assessment of (i) temporal changes of biodiversity ; (ii) during the last century ; (iii) on
148 a broad scale (at least two continents or two oceans) and (iv) at population, species and/or
149 community level. We included studies analysing temporal changes on biodiversity relying on
150 empirical data, but also reviews based on empirical assessments as well as methodological studies
151 explicitly questioning the assessment of global biodiversity changes. For studies relying on data, we
152 excluded those using exclusively human-manipulated or simulated data. We also excluded studies
153 on functional, phylogenetic and network diversity as these levels of biodiversity do not suffer from
154 the same confusions in the assessment of the changes, these were thus not the levels we aimed to
155 target. Figure 1A summarises this selection process and shows the number of articles excluded on
156 the basis of their failure to meet the above criteria. In addition, we included four reports from the
157 'Grey literature' (IPBES, 2019; IUCN, 2022; Pörtner et al., 2021; WWF, 2020), produced by
158 organisations among the best known that provide assessments of temporal changes of biodiversity
159 globally. 91 references constituted the final database, that we classified into four different
160 categories: (i) biodiversity empirical analysis (n=48) ; (ii) reviews (n=20) ; (iii) methodological
161 papers (n=19) ; (iv) reports (n=4).

162

163

164

165 ***Metadata extraction***

166 Methodological papers were considered for discussion purposes. For the other references, we
167 extracted detailed metadata and information to investigate potential sources of heterogeneities
168 regarding the Global Biodiversity Changes (Fig. 1B). We distinguished papers mostly focused on
169 biodiversity trends from those mostly focused on drivers. Papers mostly focused on drivers were
170 those considering only the effect of drivers without concluding particularly on how biodiversity
171 was changing. On the contrary, some papers were studying both trends and drivers, giving valuable
172 insights in their conclusions regarding global biodiversity changes. These were considered both for
173 analysing trends but also for analysing drivers, as we analyzed these issues separately (n=6).

174
175 For papers analyzing biodiversity trends, we recorded bibliometric data, the main conclusions, the
176 type of assessment approach, and, when relevant, information regarding the data and methods used
177 to quantify the changes (Fig. 1B). Conclusions were classified into decreasing trends, increasing
178 trends, mixed trends (i.e. there were as many increasing trends as decreasing, or a majority of no
179 trends) and factor-dependent trends (i.e. directions that varied according to certain factors, such as
180 location).

181
182 While retrieving those information, we identified two main assessment approaches adopted in the
183 corresponding studies to produce a global picture of biodiversity changes. First, a bottom-up
184 approach, which correspond to reviews or meta-analyses aggregating results of studies analyzing
185 individual datasets with varying methodologies from one study to the other (e.g. Pereira et al., 2012;
186 Pievani, 2014). Second, a top-down approach that produces a global result analyzing heterogeneous
187 data aggregated in large databases within the same methodological framework (e.g. Dornelas et al.,
188 2019; Wilson & Fox, 2021). We recorded the assessment approach for each reference and decided
189 to investigate to what extent these could explain part of the observed heterogeneous results.

190
191 When relevant and available, we also retrieved information on the data used, the ecological level
192 targeted and the methodological processes used to quantify the changes. We recorded the databases
193 used, temporal scope, taxonomic scope, and number of species. The time span was often expressed
194 through the length of the longest time series, which is why we expressed it through broad categories
195 (less than 30 years; between 30 and 50 ; more than 50 years) rather than exact numbers. The
196 taxonomic scope was entered as a list of groups among the following: mammals, birds, amphibians,
197 reptiles, fishes, invertebrates, plants (or unknown when not specified). We then categorized the
198 papers into groups regarding the number of taxa that were assessed (1 taxa; 2-3 taxa; 4 + taxa).

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

203 We categorised the methods used to quantify the changes into three main categories. The first
204 category we identified concerns papers that studied the changes through linear trends of individual
205 ecological level. These papers performed linear regressions (and all variations, e.g. state space
206 models (Daskalova et al., 2020) or logged annual growth rate (Williams et al., 2022)) on
207 biodiversity measures either population by population or species by species. These results were then
208 used to reach conclusions about the general direction of change. We will refer to these methods as
209 “linear models on individual time series”. The second category we identified corresponds to papers
210 producing global indicators built upon populations (e.g. Living Planet Index (LPI) (Loh et al.,
211 2005)) or species (e.g. Red List Index (RLI)), and estimating global trends of those single
212 aggregated metrics. The last category we identified are the other methods that were used more
213 sporadically, quantifying other aspects than the linear changes in abundance or species richness (e.g.
214 quantifying the coefficient of variation (Marsh, 2001), non-linear changes (Keith et al., 2015), or
215 other aspects like resistance and recovery (Capdevila et al., 2022)).

216

217 For papers analyzing drivers of biodiversity changes, we recorded bibliometric data, the main
218 conclusions, the type of assessment approach, and, when relevant, information regarding the
219 biodiversity data used and the identity of the studied drivers (Fig. 1B). Conclusions were classified
220 into different categories based on the nature of the impacts into negative, none or positive, and
221 factor-dependent. The assessment approaches and information regarding the biodiversity data were
222 the same as the ones described regarding the examination of the trends in biodiversity. The
223 information provided on the drivers and the methods used to quantify the drivers’ impacts on
224 biodiversity was scarce and heterogeneous. As a result, we did not address the issue of the
225 sensitivity of the results to the drivers data and methods used in depth. We rather examined the
226 impact of the identity of the drivers studied. We categorized the drivers between climate change,
227 other anthropogenic pressures (mainly land use change) and conservation policies (protected areas).

228

229 We performed chi square tests in order to test whether the conclusions drawn depended on either (1)
230 the assessment approach, (2) the dataset used, (3) the number of taxonomic groups (split in
231 qualitative categories), (4) the time span (split in qualitative categories), (5) the ecological level,

232 and (6) either the methods or the identity of the studied drivers. We performed these tests both for
233 the trends and the drivers assessments. All analyses were performed under R version 4.3.2.

234

235 **Results**

236

237 ***Global overview***

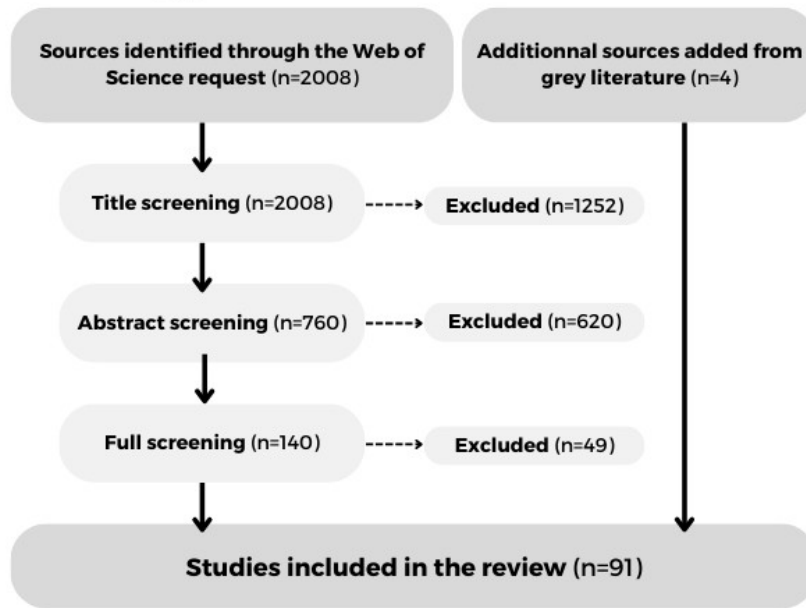
238 Interestingly, we found only 91 papers that analyze temporal changes of biodiversity globally, while
239 a plethora of examples accumulate at more local scales (e.g. Donald et al., 2001; Koleček et al.,
240 2021). These were published from 1991 to 2021, with a majority published after 2010 (Appendix
241 S2). We identified an increasing interest in this question over time, although this tendency also
242 reflects an overall increase in the number of papers published during the same period. Of the 91
243 papers, 48 directly relied on empirical datasets, 20 were reviews, 19 were methodological papers
244 and 4 were reports that we added from grey literature.

245

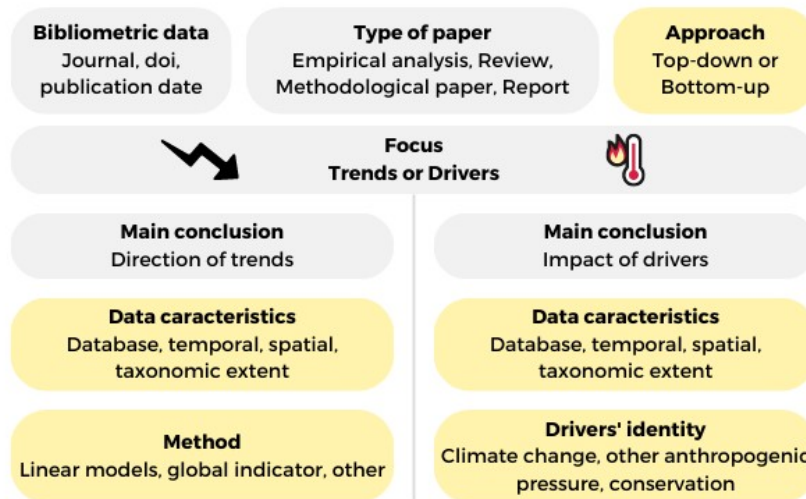
246 Out of the 44 papers assessing temporal trends in biodiversity changes, 57% (n=25) concluded that
247 biodiversity is globally in decline. More than a third of the papers concluded that trends are mixed
248 or factor-dependent (36%, n=16). Only 3 of the reviewed papers showed evidence of increasing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

253

A Screening process



B Metadata extraction



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

259 **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 ($n_{calc}=21$)

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

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

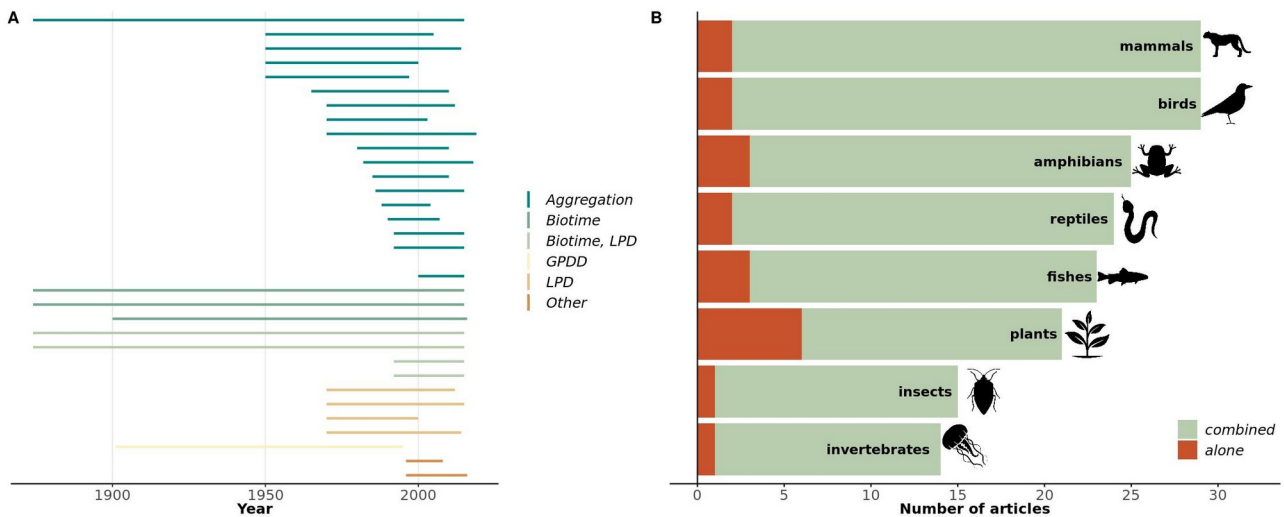
265 ** Aggregation of several locally available data sets in individual studies.

266

267 Papers relying on empirical data in majority used aggregations of several data sets from individual
 268 studies (62,5%, $n=30$). Still, we found that three main global databases gathering biodiversity data
 269 were used: BioTIME (Dornelas et al., 2018), the Living Planet Database (LPD) (Loh et al., 2005),
 270 and the Global Population Dynamics Database (GPDD) (Inchausti & Halley, 2001). The remaining
 271 papers relied on other unique but more sporadically used databases that were not open access. Table
 272 1 highlights the characteristics and properties of the different data sources. The overwhelming
 273 majority of papers were based on terrestrial data covering all continents (67%, $n=32$), which is not
 274 surprising since our selection process was based in part on the choice to have global studies. Still,
 275 we accepted articles covering at least two continents, and we note that the only continents that were
 276 studied in pair were America and Europe. The temporal coverage of the data was quite

277 heterogeneous, but mainly covered the last 70 years (Fig. 2A). This also confirms the strength of
 278 BioTIME's temporal coverage compared to other databases (Dornelas et al., 2018). The taxonomic
 279 coverage shows that the papers in our corpus mainly relied on data referring to several groups (Fig.
 280 2B). As well as for the geographical extent, this is not surprising considering our selection process.
 281 Still, there is a preponderance of data on mammals and birds, more generally on terrestrial
 282 vertebrates, and an under-representation of insects and invertebrates.
 283 The empirical data were based on different levels of biodiversity (Fig. 2C). Papers relying on
 284 biodiversity at the population scale (e.g. abundances, biomass, density) represented 50% of our
 285 corpus (n=24), papers relying on biodiversity at the species scale (e.g. species richness, detection
 286 rates) represented 31% (n=15). The 19% (n=9) remaining relied on grouped levels, i.e. several
 287 levels among population, species, or also community scale (studied through community
 288 composition).

289



290

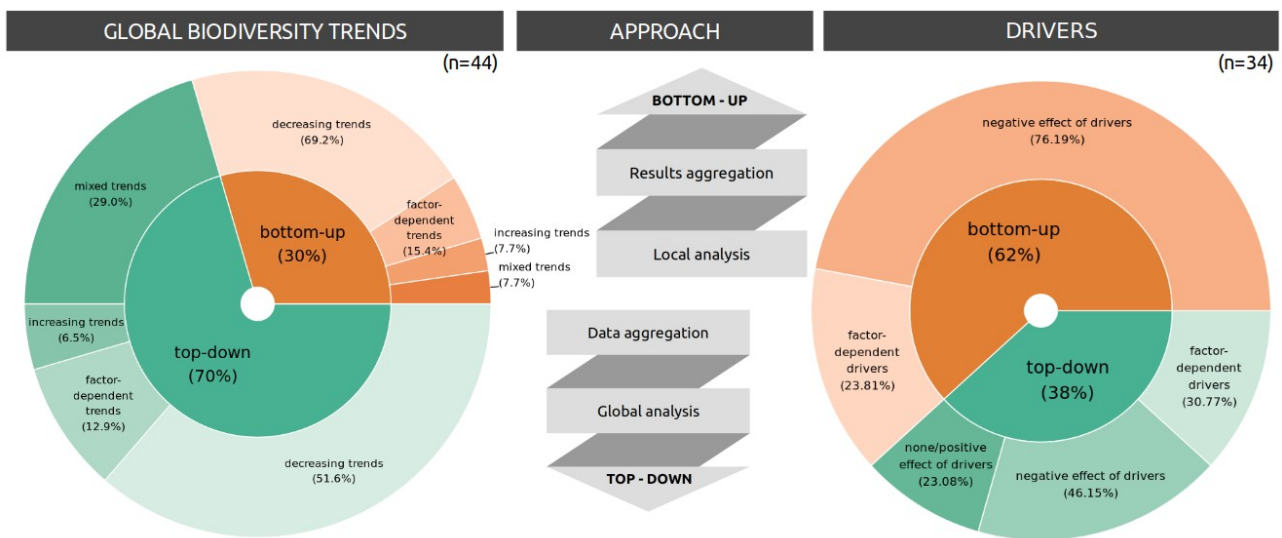
291 **Figure 2. Information on the biodiversity data on which the reviewed papers are based.** (A)
 292 Temporal extent of the data, colored based on the data source (see Table 1). (B) Taxonomic extent
 293 of the data. Number of papers considering each group, whether considered alone or combined (i.e.
 294 when the study does consider several groups at a time). For instance, 2 papers were examining the
 295 fate of mammals only and 30 were examining the fate of mammals and other groups at the same
 296 time.

297 In the following, we describe how studies of global biodiversity change are distributed among
 298 several criteria. We performed formal statistical analysis using chi square tests to see how papers
 299 were distributed among groups. Only two of the tests were significant (the one regarding the impact
 300 of the assessment approach on the drivers impacts and the one looking at the identity of the drivers,
 301 see appendix S4 for details). However, the low amount of papers available for each single analysis
 302 call for cautious interpretations of these tests (see discussion).

303

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

305 We investigated three potential sources of heterogeneities. First, we examined the effect of using
 306 different assessment approaches. When papers used a bottom-up approach (i.e. aggregating local or
 307 regional results with varying methodologies into meta-analyses or reviews), they mostly concluded
 308 a decline in global biodiversity (69%, Fig. 3). In contrast, when papers used a top-down approach
 309 (i.e. performing analyses through a single methodology on global datasets), their conclusions were
 310 much more balanced: a bit more than 50% concluded that biodiversity was decreasing and almost
 311 50% that trends were mixed or increasing (Fig. 3).



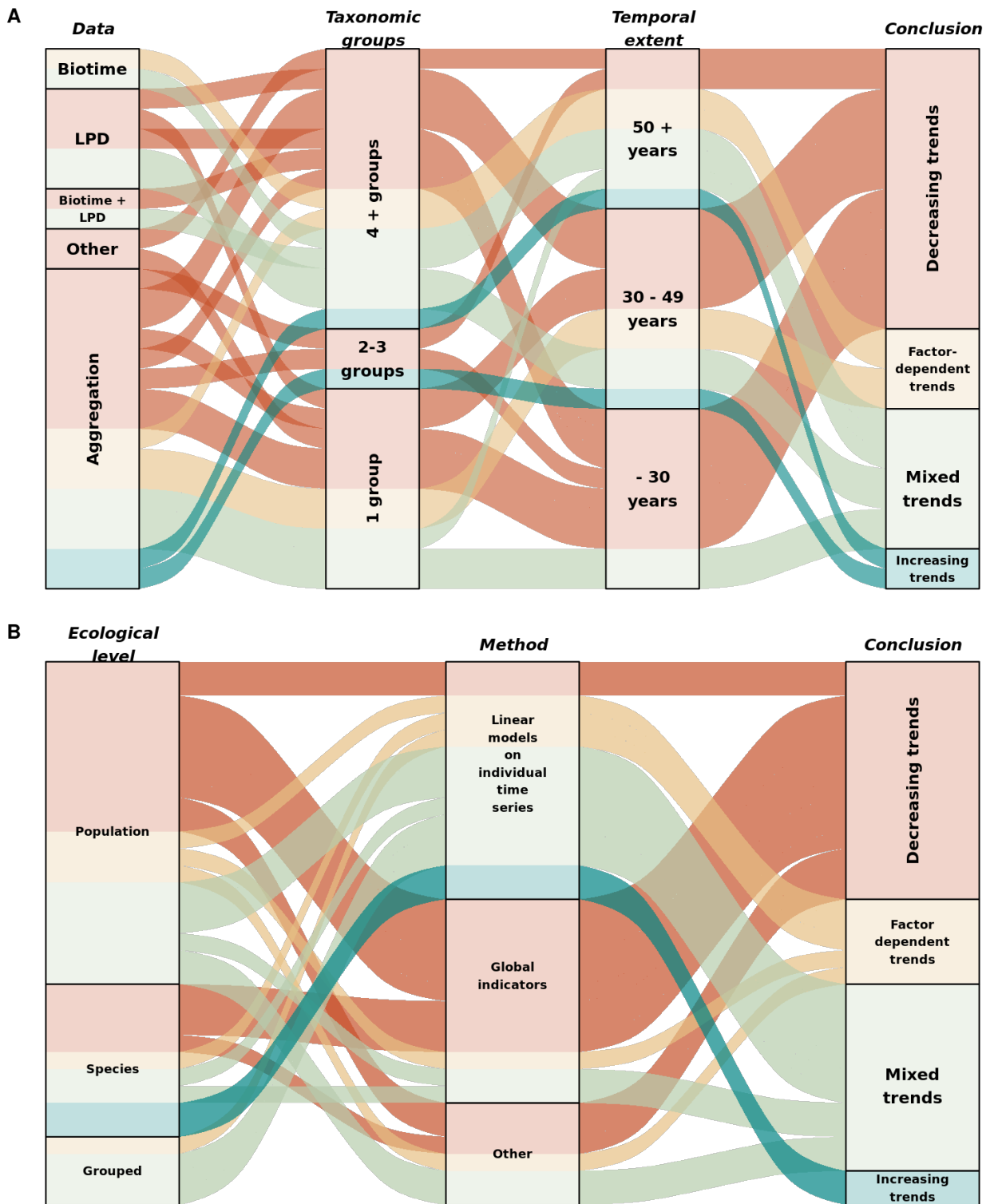
312 **Figure 3. Proportions within conclusions depending on whether global biodiversity trends or**
 313 **drivers were assessed and depending on the assessment approach (i.e. top-down or bottom-**
 314 **up).**

315

316 Second, we examined the effect of using data with different characteristics. Among the 22 papers
 317 using aggregated databases, 45% concluded a decline, 41% mixed or factor-dependent trends and
 318 14% identified increases in biodiversity globally. The LPD and the other individual datasets were
 319 mainly associated with a decline in global biodiversity, whereas studies using the BioTIME

320 database mainly concluded that trends were mixed or factor-dependent (Fig. 4A). The conclusions
321 between declining or mixed trends were balanced no matter the number of taxonomic groups
322 considered (Fig. 4A). The time span seemed to be influencing the conclusions however. The longer
323 the data, the more heterogeneous the results were: 78% of articles based on time series of less than
324 30 years concluded that trends were declining, whereas this percentage dropped to 60% for time
325 series between 30 and 49 years long, and to 22% for time series longer than 50 years.

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



337

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

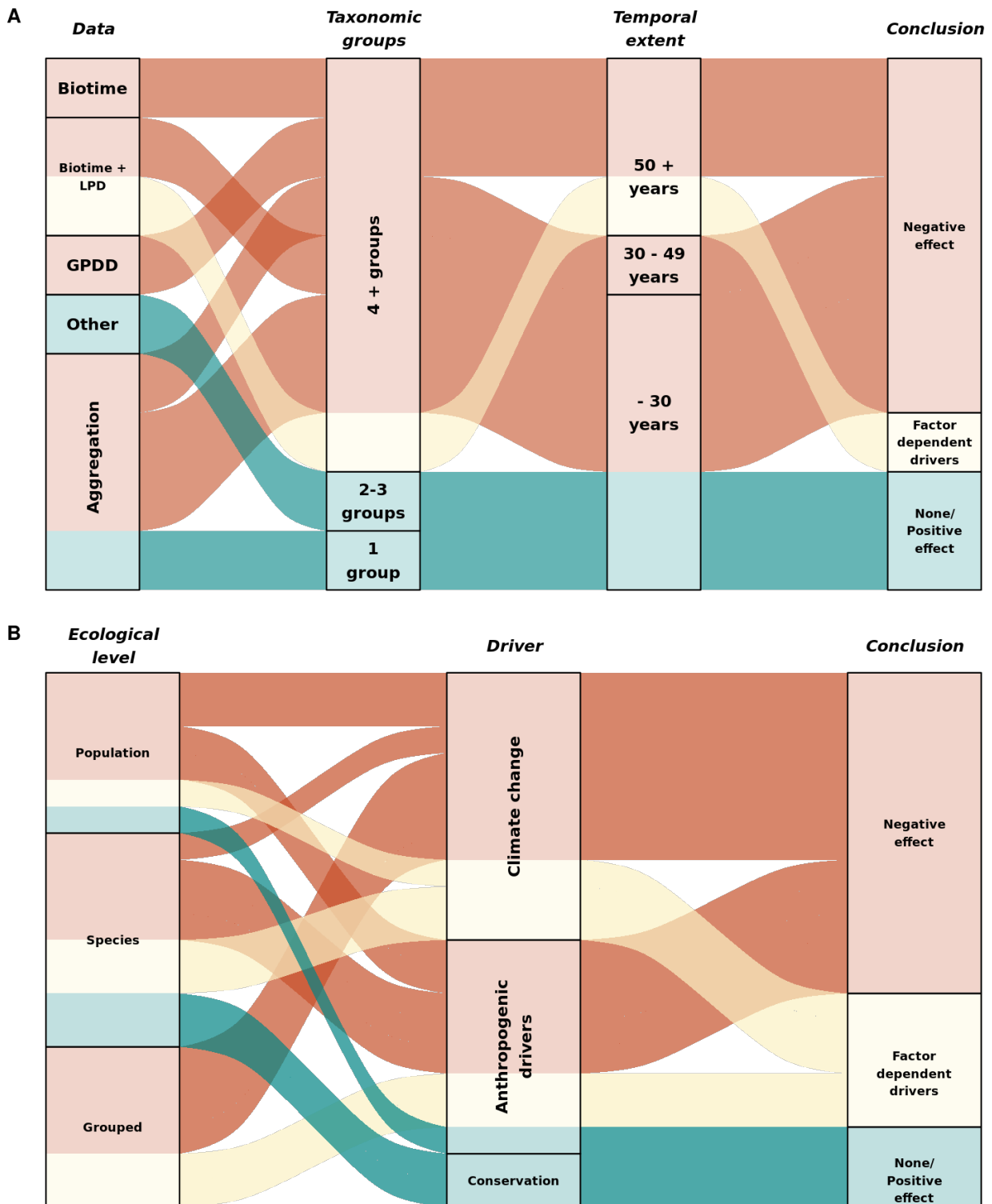
343

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

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

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

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



362
 363 **Figure 5. Alluvial plots representing the links between the potential sources of heterogeneity**
 364 **and the conclusions when assessing the impact of the drivers.** A (n=9) highlights the impact of
 365 the data characteristics. B (n=20) highlights the impact of the biodiversity level and the identity of
 366 the drivers. The height of the boxes is proportional to the number of papers. Flow thickness between
 367 boxes is proportional to the number of papers. Colours of the flow reflect the conclusions.

368

369

370 **Discussion**

371 Quantification of temporal changes in biodiversity at large scales and attribution of drivers of these
372 changes is a daunting task. Studies quantifying global declines are mixed with evidences of more
373 heterogeneous changes, including declines, increases, and no net change at different levels of
374 biological diversity. In a context of accelerated global change, clarifying and addressing these
375 sources of heterogeneity in temporal changes of biodiversity is needed to inform conservation
376 policies. Plus, far from weakening the knowledge on biodiversity loss, working on uncertainty is in
377 fact the best way to consolidate what we know. Here, we explored how different methodological
378 pathways to produce estimates of biodiversity changes likely influence the direction of trends in
379 population abundance, species richness and community composition as well as the effect of the
380 drivers of these trends. As our analysis is not based on a systematic review, we recognize that the
381 coverage of the literature is probably not complete. Moreover, with the explosion of open databases
382 worldwide, articles on this subject are accumulating very quickly (Appendix S2). Therefore, there
383 might be a gap between the conclusions drawn by the corpus considered in this study and what will
384 possibly be reflected by this very active field in the near future. However, the robustness of our
385 methodology to target a broad range of papers within the initial query allows a certain degree of
386 confidence regarding the conclusions we may currently draw in the present paper.

387 The number of papers found was quite surprisingly low and did not allow any advanced statistical
388 analysis. However, the present review is not meant to be systematic nor to generate quantitative
389 estimates, but is primarily conceptual. Yet, the non-significance of the chi-squared tests confirm the
390 absence of dominating type of studies addressing global biodiversity changes (although the low
391 sample sizes deserve cautious interpretation for some of the comparisons). But beyond figures and
392 tests, we demonstrate that part of the confusion revolving around global biodiversity changes is to
393 find in the diversity of definitions, methods and approaches adopted to address this issue.

394

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

396 Producing biodiversity syntheses requires gathering empirical evidence of biodiversity changes
397 distributed on a broad scale. We have identified two major ways of estimating global changes in
398 biodiversity: either by synthesizing already available information (“bottom-up”) or by producing
399 estimates from raw global data (“top-down”). Our findings show that the bottom-up syntheses from
400 reviews or meta-analyses most often conclude a decline in biodiversity. This could be due to biases
401 in the selected studies when performing such bottom-up assessment. The political intent of
402 governments or conservationists to monitor more endangered species results in a selection bias if
403 the trend of those species is interpreted as an average trend for the entire taxonomic group

404 considered (Boakes et al., 2010). Such selection bias might be further amplified due to a publication
405 bias: as studies generally hypothesize a decline, any study showing neutral or positive change
406 would fail to prove the hypothesis, which may encourage authors to select more results on declining
407 trends (Haddaway et al., 2020; Mlinarić et al., 2017). Studies showing biodiversity declines are thus
408 over-represented compared to studies showing neutral or positive changes. This tendency is
409 exacerbated by bottom-up assessments as they rely on already published materials. We do not imply
410 that reviews or meta-analyses should not be conducted, but rather that biodiversity syntheses should
411 make sure publication bias is taken into account (see Haddaway et al., 2020 for recommendations).

412

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

414 Quantification of trends is also based on empirical biodiversity data that are biased in terms of
415 spatial coverage and taxa considered. Plus, these data often represent monitoring of populations or
416 communities through short sampling periods, hence the challenges around the use and collection of
417 these data. Our findings confirm the major geographical and taxonomic drawbacks when estimates
418 are generated from empirical data, namely geographical and taxonomic ones. Europe and North
419 America are the most assessed continents (Boakes et al., 2010; Manes et al., 2021; Saha et al.,
420 2018), whereas information is least available in the tropics (Feeley & Silman, 2011; Saha et al.,
421 2018) and in regions that are currently under pressure (Pereira et al., 2012). Regarding taxonomic
422 groups, terrestrial organisms, and especially vertebrates (Davison et al., 2021; Pereira et al., 2012;
423 Theobald et al., 2015), are more studied than marine ones (Manes et al., 2021). Overall, there is an
424 over-representation of endangered species (Boakes et al., 2010; Saha et al., 2018). In addition to
425 these already well-documented taxonomic and geographical biases, we show that the length of the
426 period considered is influencing the conclusions: the reviewed studies that are based on short time
427 series identify more declines. If species historically monitored are recovering, short time series may
428 miss their recovery. Selecting longer time-series should buffer the decline in this scenario. These
429 results contradict other findings though. For instance, Vellend et al. (2013) found that the length of
430 the time series had no effect on the assessment of biodiversity change. On the contrary, Gonzalez et
431 al. (2016) showed that the percentage of decline increased with the length of the time series.

432

433 Thus, the heterogeneity in the conclusions is probably related to heterogeneity in the data
434 characteristics. Beyond being a simple source of heterogeneity, this indicates a lack of
435 representativeness that may represent a bias and therefore influence the reliability of the estimated
436 trends. The lack of representativity in certain groups or regions does not ensure the reliability of the
437 trends, but recent studies highlighted that accounting for these biases (e.g. through weighting

438 processes (McRae et al., 2017)) is not sufficient to correctly assess the trends (Dove et al., 2023).
439 They showed that not only short time series are less reliable than longer ones (Wauchope et al.,
440 2019), but also that the assessment of temporal changes in biodiversity at the global scale depends
441 more on the number of time series (considered here at population scale) than on the
442 representativeness of the number of species present within each taxonomic group in the data (Dove
443 et al., 2023).

444

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

452

453 ***Beyond linear trends***

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

459

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

465

466 Our results show that linear models applied to individual time series seem to be a less biased
467 modelling strategy compared to global indicators. Linear models are often used to estimate the rate
468 of change of a variable (Christensen et al., 2014; Donald et al., 2001; Sánchez-Bayo & Wyckhuys,
469 2019), the most widely used variable being species richness (Hillebrand et al., 2018). However,
470 species richness may not be the most appropriate variable to measure biodiversity change. Indeed,
471 beyond being widely used, changes in species richness remain poorly informative. It has been

472 proven to be insensitive to other form of biodiversity change (Hillebrand et al., 2018; Santini et al.,
473 2017) and unreliable to detect direction in trends despite the relative simplicity to calculate it
474 (Valdez et al., 2023).

475

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

486

487 Also, most attention is given on trends while other characteristics in the pattern of biodiversity
488 changes are most often ignored. Variability (and changes thereof) is a proxy of stability in
489 ecological systems (Donohue et al., 2016). Yet, the variability in biodiversity dynamics is largely
490 overlooked: variability is often studied at the scale of ecosystems or communities to characterize
491 their stability (Hughes et al., 2013; Scheffer et al., 2009), but very little at the scale of populations
492 in the context of a global analysis. In the studies we reviewed, stability has only been investigated
493 once (Marsh, 2001), and the few examples of studies that have assessed population stability at the
494 global scale involve taxonomically and spatially biased data (Leung et al., 2017; Williams et al.,
495 2022). Instead, most studies focus on the extinction of species (Bellard et al., 2012) and ignore
496 fluctuations in populations, although fluctuations can indicate high vulnerability to extinction
497 (Clements et al., 2015).

498

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

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

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

517

518 ***Conservation perspectives***

519 Although some of the biases we report here potentially lead to over or under estimate the overall
520 decline in biodiversity, we do not question the magnitude of the biodiversity crisis. We are now
521 moving forward with the post-2020 Global Biodiversity Framework and all the Aichi Targets for
522 2020 have been only partially achieved or not achieved at all. New goals have been set within the
523 Kunming-Montreal Global Biodiversity Framework, many of which rely on the way biodiversity is
524 perceived to be changing. The Intergovernmental Platform on Biodiversity and Ecosystem Services
525 (IPBES) has published a very alarming global biodiversity assessment report in 2019. In this
526 context, characterizing the state of biodiversity, the impact of drivers and responses is a key step to
527 take action and “bend the curve of biodiversity loss” (Tekwa et al., 2023). While we urgently need
528 reliable assessments to quantify temporal changes in biodiversity and their links to global drivers,
529 we call for more attention to overlooked and yet informative components of biodiversity changes.

530

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

538

539 We suggest that conservation actions should be based not only on Essential Biodiversity Variables
540 but more globally on Essential Biodiversity Data, with requirements on taxonomic, geographic and
541 temporal coverage ensuring the reliability of estimated trends and thus being able to guide strategies
542 based on the least biased observations possible. Similarly, the implementation of a framework such
543 as Essential Biodiversity Assessment Methods, including the need to systematically take into
544 account different levels of biodiversity, as well as the measurement not only of linear dynamics,
545 should be considered collectively and on a large scale.

546

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

555

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

569

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

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

573 **Conflict of interest disclosure**

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

576

577 **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>
- 578 Díaz, S., & Malhi, Y. (2022). Biodiversity : Concepts, Patterns, Trends, and Perspectives. *Annual*
579 *Review of Environment and Resources*, 47(1), 31-63. [https://doi.org/10.1146/annurev-
environ-120120-054300](https://doi.org/10.1146/annurev-
580 environ-120120-054300)
- 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.

- 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>
- 581 Mammola, S., Fukushima, C. S., Biondo, G., Bongiorno, L., Cianferoni, F., Domenici, P., Fruciano,
582 C., Lo Giudice, A., Macías-Hernández, N., Malumbres-Olarte, J., Miličić, M., Morganti, M.,
583 Mori, E., Munévar, A., Pollegioni, P., Rosati, I., Tenan, S., Urbano-Tenorio, F., Fontaneto,
584 D., & Cardoso, P. (2023). How much biodiversity is concealed in the word ‘biodiversity’?
585 *Current Biology*, 33(2), R59-R60. <https://doi.org/10.1016/j.cub.2022.12.003>
- 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.0604181103>
- 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., Ceaușu, 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.