

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

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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 assessing such changes is of major social, political, and scientific importance. Part of this
12 variability may arise from the difficulty to reliably assess global biodiversity trends. Here, we
13 conducted a literature review of studies documenting the temporal dynamics of global biodiversity.
14 We classified the differences among approaches, data and methodology used by the reviewed
15 papers to reveal common findings and sources of discrepancies. We show that reviews and meta-
16 analyses, along with the use of global indicators, are more likely to conclude that trends are
17 declining. On the other hand, the longer the data are available, the more nuanced are the trends they
18 generate. Our results also highlight the lack of studies providing information on the impact of
19 synergistic pressures on a global scale, making it even more difficult to understand the driving
20 factors of the observed changes and how to decide conservation plan accordingly. Finally, we stress
21 the importance of taking into account the sources of confusion identified, as well as the complexity
22 of biodiversity changes, in order to implement effective conservation strategies. In particular,
23 biodiversity dynamics are almost systematically assumed to be linear, while non-linear trends are
24 largely neglected. Clarifying the sources of confusion in global biodiversity trends should
25 strengthen large scale biodiversity 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 (Díaz & Malhi, 2022;
63 Mammola et al., 2023). For example, the WWF considers biodiversity at the population level
64 (WWF, 2020, 2022). In contrast, others (e.g. Dornelas et al., 2014) consider biodiversity at the
65 species or community level. As declines in population sizes and species richness are not necessarily
66 related, both an increase and decrease in "biodiversity" can be concluded depending on the
67 ecological level of interest. Clarifying the trends for each level of biodiversity should, in principle,
68 limit the confusion. In practice however, trends within the same ecological levels show both a
69 decline globally or no net changes; either at species scale (IUCN, 2019; Dornelas et al., 2014;
70 Vellend et al., 2013) or at population scale (Daskalova et al., 2020; He et al., 2019; Leung et al.,
71 2020; Wagner et al., 2021). Beyond the ecological level considered other factors are therefore also
72 generating confusion in the observed results.

73

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

90

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

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

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

117

118 **Methods**

119

120 ***Literature review***

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

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

141

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

159

160

161

162 ***Metadata extraction***

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

171

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

178

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

187

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

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

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

213

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

225

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

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

231

232 **Results**

233

234 *Global overview*

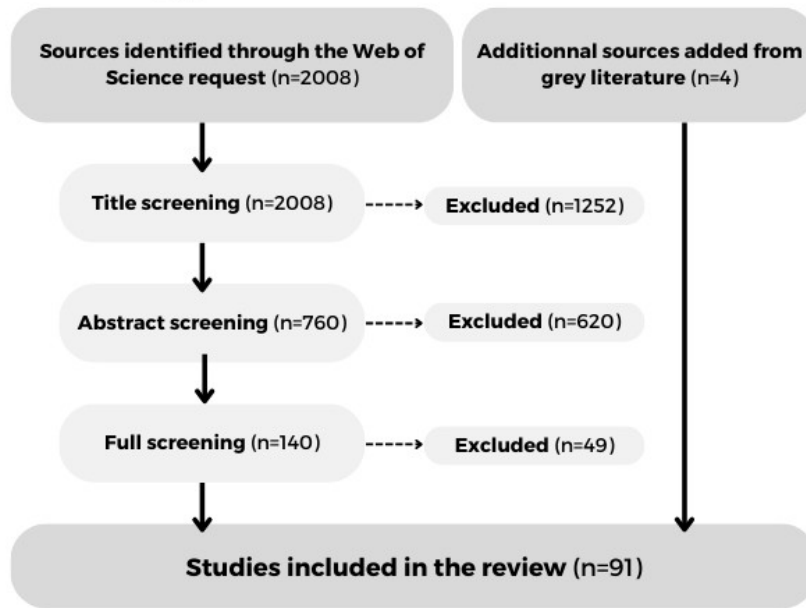
235 Interestingly, we found only 91 papers that analyze temporal changes of biodiversity globally, while
236 a plethora of examples accumulate at more local scales (e.g. Donald et al., 2001; Koleček et al.,
237 2021). These were published from 1991 to 2021, with a majority published after 2010 (Appendix
238 S2). We identified an increasing interest in this question over time, although this tendency also
239 reflects an overall increase in the number of papers published during the same period. Of the 91
240 papers, 48 directly relied on empirical datasets, 20 were reviews, 19 were methodological papers
241 and 4 were reports that we added from grey literature.

242

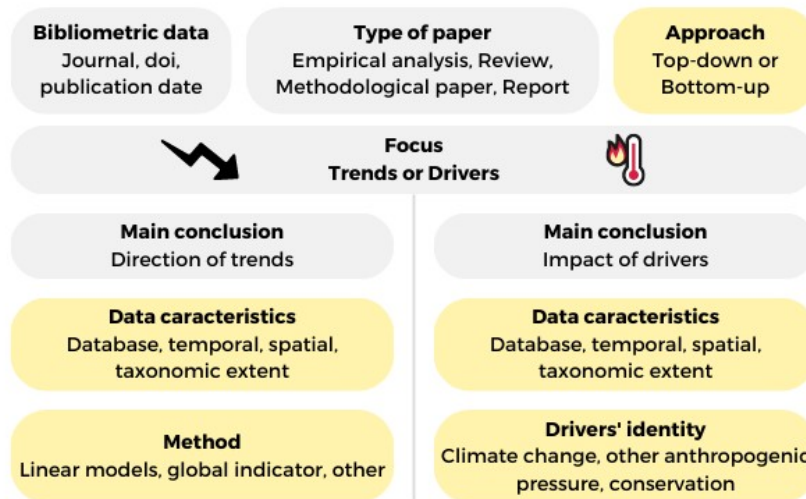
243 Out of the 44 papers assessing temporal trends in biodiversity changes, 57% (n=25) concluded that
244 biodiversity is globally in decline. More than a third of the papers concluded that trends are mixed
245 or factor-dependent (36%, n=16). Only 3 of the reviewed papers showed evidence of increasing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

250

A Screening process



B Metadata extraction



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

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

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

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

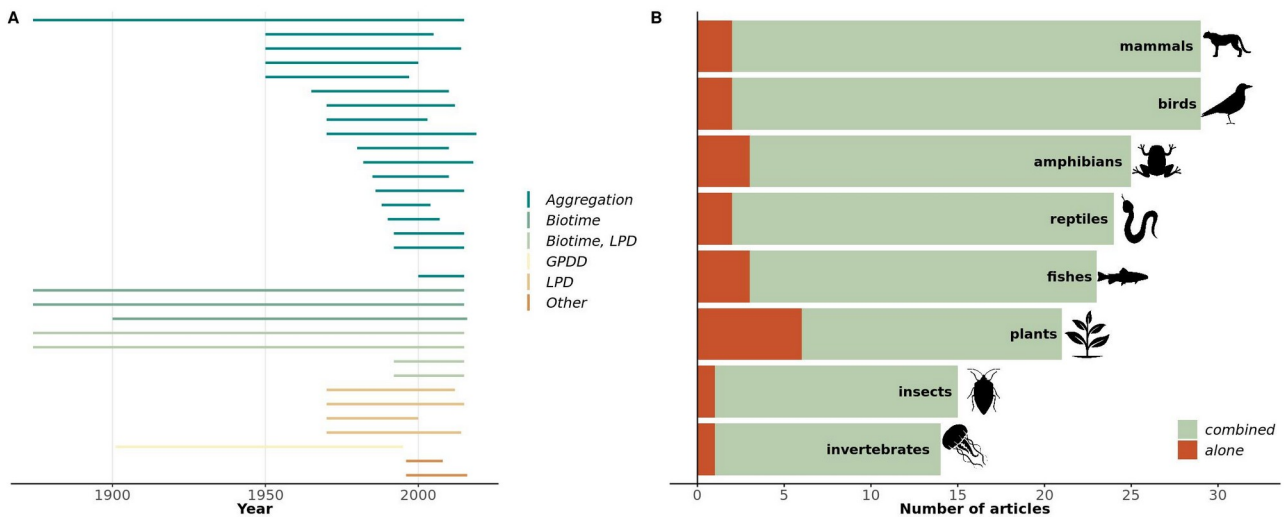
262 ** Aggregation of several locally available data sets in individual studies.

263

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

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

286



287

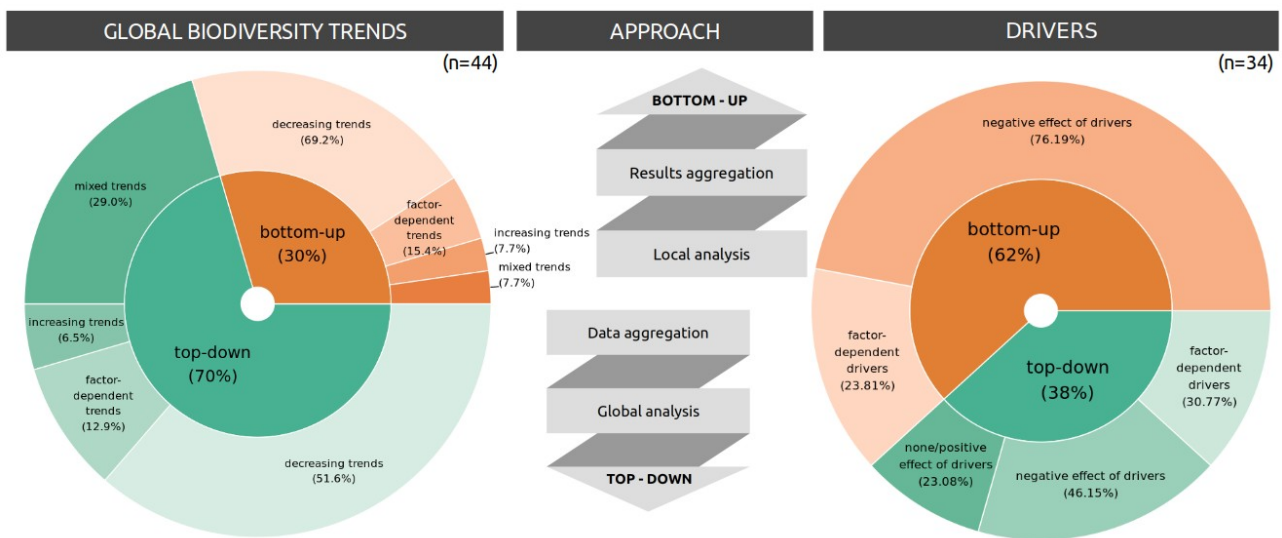
288 **Figure 2. Information on the biodiversity data on which the reviewed papers are based.** (A)
 289 Temporal extent of the data, colored based on the data source (see Table 1). (B) Taxonomic extent
 290 of the data. Number of papers considering each group, whether considered alone or combined (i.e.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

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

300

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

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



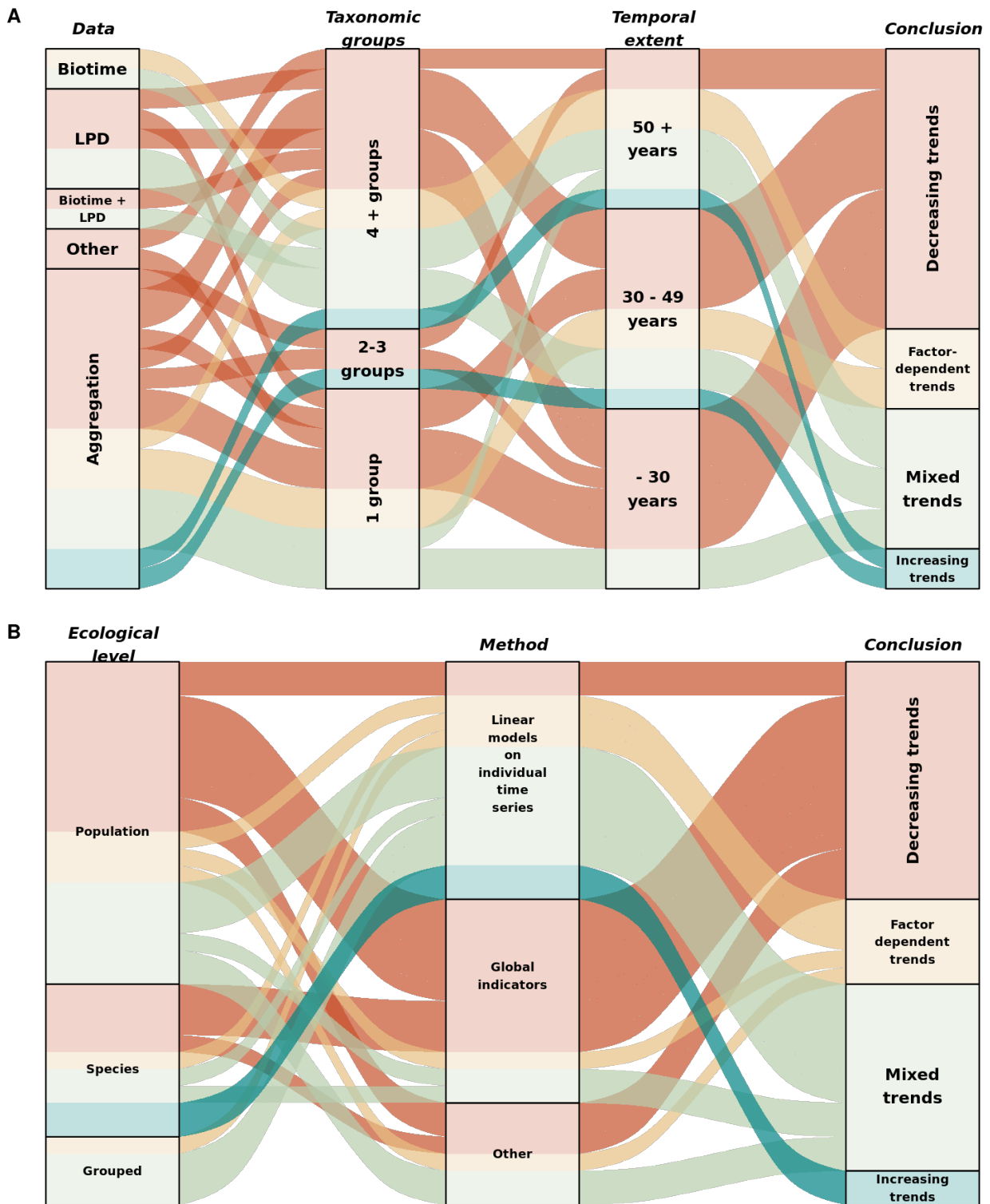
309 **Figure 3. Proportions within conclusions depending on whether global biodiversity trends or**
 310 **drivers were assessed and depending on the assessment approach (i.e. top-down or bottom-**
 311 **up).**

312

313 Second, we examined the effect of using data with different characteristics. Among the 22 papers
 314 using aggregated databases, 45% concluded a decline, 41% mixed or factor-dependent trends and
 315 14% identified increases in biodiversity globally. The LPD and the other individual datasets were
 316 mainly associated with a decline in global biodiversity, whereas studies using the BioTIME

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

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



334

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

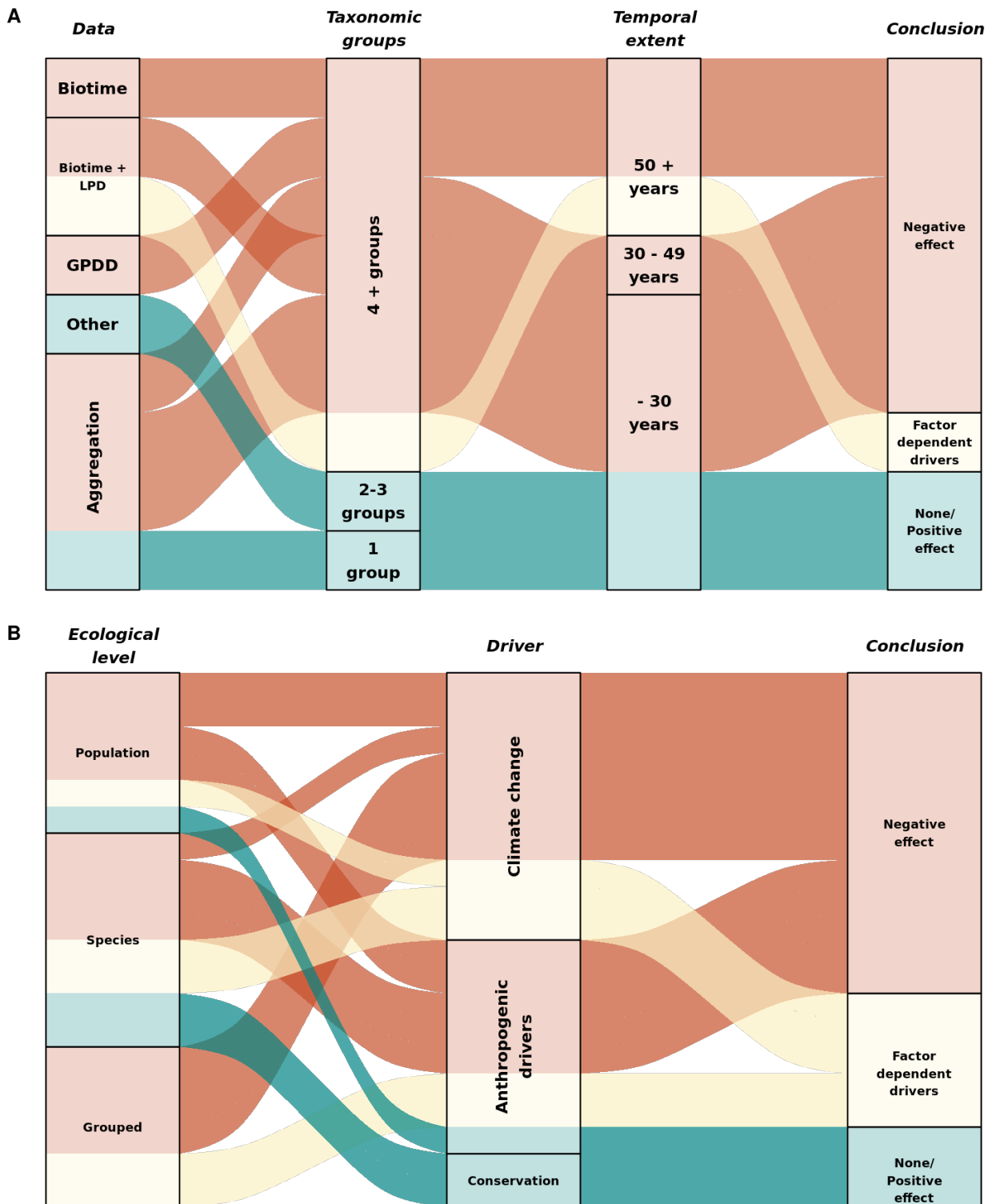
340

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

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

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

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



359
 360 **Figure 5. Alluvial plots representing the links between the potential sources of heterogeneity**
 361 **and the conclusions when assessing the impact of the drivers.** A (n=9) highlights the impact of
 362 the data characteristics. B (n=20) highlights the impact of the biodiversity level and the identity of
 363 the drivers. The height of the boxes is proportional to the number of papers. Flow thickness between
 364 boxes is proportional to the number of papers. Colours of the flow reflect the conclusions.

365

366

367 **Discussion**

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

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

391

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

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

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

409

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

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

429

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

435 processes (McRae et al., 2017)) is not sufficient to correctly assess the trends (Dove et al., 2023).
436 They showed that not only short time series are less reliable than longer ones (Wauchope et al.,
437 2019), but also that the assessment of temporal changes in biodiversity at the global scale depends
438 more on the number of time series (considered here at population scale) than on the
439 representativeness of the number of species present within each taxonomic group in the data (Dove
440 et al., 2023).

441

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

449

450 ***Beyond linear trends***

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

456

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

462

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

469 proven to be insensitive to other form of biodiversity change (Hillebrand et al., 2018; Santini et al.,
470 2017) and unreliable to detect direction in trends despite the relative simplicity to calculate it
471 (Valdez et al., 2023).

472

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

483

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

495

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

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

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

514

515 ***Conservation perspectives***

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

527

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

535

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

543

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

552

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

566

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

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

570 **Conflict of interest disclosure**

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

573

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