



Literature review

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"Scaling Biodiversity Trends"

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1. Introduction

Our life quality is intrinsically linked to the state of ecosystems that we live in. Ecosystem services to humans (Diaz et al. 2018), as well as their inner functioning, involve a spectrum of mechanisms including nutrient cycles or ecosystem stability (Pereira, Navarro, and Martins 2012). Some of these functions, such as seed dispersal, control of pests, scavenging, or pollination, depend on birds and their diversity. Birds also have a significant esthetical value, with most countries having numerous iconic/charismatic species of conservation interest (<https://www.iucnredlist.org/>). Moreover, given their ability to quickly move between locations, their presence is also a good indicator for ecosystem health. Unfortunately, anthropogenic stressors like habitat loss, over-exploitation, pollution, or introduction of invasive species are a threat to birds and their biodiversity (Donald et al. 2001, Jiguet et al. 2010), with concerns that they could face a sixth mass extinction (Barnosky et al. 2011).

We have reasons to suspect that the global alteration of biodiversity due to anthropogenic stressors is unprecedented, and political goals have been declared in order to limit it (e.g. Convention on Biological Diversity and Centre 2006, 2010, 2002). However, data-driven basis for these policies remains a challenge, mainly due to severe gaps and biases in empirical biodiversity data (Meyer et al. 2015). To complicate matters further, current scientific literature has shown that temporal trends of local biodiversity can be opposite to trends at larger spatial scales (e.g. Keil et al. 2011, 2018, Chase et al. 2019). Thus, we should expect changes in biodiversity to be far more complex than a simple global decrease. Finally, biodiversity can be measured by many metrics, and these can differ in their temporal trends (McGill et al. 2015); for instance, while there may be small average net change in local species richness, ecosystems can still undergo significant changes in species composition (Blowes et al. 2019; Dornelas et al. 2013; Vaidyanathan 2021).

Particularly the issue of scale is critical (Levin 1992). Since Arrhenius (1921) and Grinnell (1922), we know that spatial and temporal scaling of biodiversity affects macroecological patterns. Even though particularly the static spatial scaling of biodiversity has been of great interest (e.g. Storch, Marquet, and Brown 2007), wonders persist about how temporal trends of biodiversity are linked to the spatial and temporal scales. In other words: does the observed biodiversity trends differ if we zoom out from local communities to regions, countries or continents? Here, the term “spatial grain” is also used to refer to the spatial scale of biodiversity, i.e. the area at which the metric of biodiversity is assessed. One should be

careful to not confuse spatial grain with the spatial extent of a study, i.e. the total area of the ecosystem which is observed or analyzed (Dungan et al. 2002). The same terminology is applied for the temporal scale: temporal grain refers to the temporal unit of the measured biodiversity (Adler and Lauenroth, 2003). Compared to spatial scaling, the temporal scaling is much less studied, which is mostly due to the lack of temporally replicated data. This review will, in part, show how the definition of temporal grain has still no consensus in the scientific literature.

To investigate the link between spatio-temporal grains and trends of biodiversity, birds are a relevant taxon. Thanks to the many ornithological monitoring initiatives and surveys, we have a large number of high-quality time series on bird populations (Bejček and Stastný 2016; Sauer et al. 2013; Kamp et al. 2021...). This is because birds are easy to observe, easy to identify and thus many volunteers are motivated to participate on citizen-science projects (cite eBird, iNaturalist) or to conduct standardized sampling (e.g. most of the breeding bird surveys are conducted by volunteers).

Given birds importance to assess ecosystem health, several standardized metrics have been created to assess their populations. Fraixedas et al. (2020) reviewed this wide spectrum of bird biodiversity indicators, without considerations to their link with spatial and temporal grains. Here, we review articles assessing the temporal trends of avian biodiversity, with species focus on the variety of metrics that they use, and spatial and temporal scales at which temporal trends have been assessed. We consider the most common macroecological indicators used to assess biodiversity at the community level and higher, such as diversity indexes (e.g. species richness, functional diversity) or population indexes (McGill et al. 2015). We highlight a lack of consensus about specifications and definitions of both spatial and temporal grains (respectively) of trends, we demonstrate that the scaling of trends is seldom considered, leading to confused messages about overall trends. Moreover, we show that studies lack spatial replication that would make reported trends robust and general. We believe that this review will improve the current knowledge on spatio-temporal scaling of trends biodiversity, and thus be useful for the ornithological field community and the conservation decision making.

2. Metrics and indicators

Studying biodiversity can be confusing as it exists many ways to measure it. The different metrics and indicators have diverse features, and one should consider which one is the most suited to its study. First, the type of biodiversity studied must be chosen (e.g. taxonomic, functional, phylogenetic diversity). Only

then, one must choose the metric(s).

Classical metrics. Measures of static biodiversity are commonly used such as local species richness (α diversity), regional richness (γ diversity) (Whittaker 1960), by indices that consider abundances (e.g. Shannon 1948, Simpson 1949), or by Hill numbers (Hill 1973). On the other hand, change of species composition in space and time can be expressed as $\beta = \frac{\gamma}{\alpha}$ (Whittaker 1972), or by pairwise dissimilarity among locations or time periods (Koleff et al. 2003). All these metrics assess species-based metrics, i.e. they use the species as a unit. However, it has also been shown that functional and phylogenetic diversity can provide supplementary information on the community structure and its dynamic (e.g. McGill et al. 2006; Mouquet et al. 2012; Webb et al. 2002).

Composite and multi-species indicators. The composite indicators are made to summarize several ecosystem information into one informative index. The most known ones are the Red List Index (Stuart H. M. Butchart et al. 2007; S. h. m. Butchart et al. 2005; Stuart H. M. Butchart et al. 2004), the Living Planet Index (Lohl et al. 2005) or the Biodiversity Change Index (Normander et al. 2012). In these composite indicators, metrics of great interest are the abundance-based or population-based metrics. As individuals react to stress or disturbances, the population trends reflect ecosystems health. The population decline that a species undergoes before going locally extinct is not captured by species-based metrics. Thus, population trends are usually efficient at assessing finer biodiversity declines. Although overall abundance is often hard to assess, the abundance of few indicator species can reflect processes in an entire ecosystem (Richard D. Gregory et al. 2005). This has led to a proposition of a family of metrics called the multi-species indicators (MSI, Landers et al 1988). Examples are the farmland bird indicator, woodland bird indicator or Wildland Bird Indicator which summarizes the latter two (Richard D. Gregory et al. 1999; Richard D. Gregory and Strien 2010). These metrics compute the geometric mean of abundance of few key species over time.

3. Quantitative literature review

For this review, we focused on articles assessing temporal trends of the most common metrics of avian biodiversity and specifying spatial and temporal scales, which are at the same time consistent with broader macroecological scaling and theory (Storch et al. 2007, McGill et al. 2015), and thus can be compared across studies and scales. Namely, these were: *Species richness*, *Evenness*, *Abundance*, *Diversity*, *Temporal beta-diversity*, *Spatial beta-diversity*, *Functional diversity*, *Functional evenness*, *Functional richness*. Some

of these classes contain several different indexes. For instance, the class Diversity, which contains either the Shannon or Simpson index, or the class Abundance, which contains various multi-species indicators (see Table 6.1).

We only considered articles for which there were spatial replicates, i.e. where the trend of the metric was assessed at several locations at a given spatial grain. With these replications, the trend reported at one spatial grain is more reliable and general. However, at larger spatial grains (i.e. national, continental or global scales), spatial replicates are rare. Thus, we considered these trends from a single location only when based on a large set of observations from smaller spatial scales.

We used the “advanced search” tool of the ISI Web of Science Core collection database with these four following queries:

1. AB = ((biodiversity OR species richness OR diversity) AND (temporal trend* OR dynamic*) AND (bird* OR avia*)) which resulted in 1346 references.
2. AB = ((biodiversity change index) AND (bird* OR avia*) AND trend*) which resulted in 60 references.
3. AB = ((species richness) AND (bird* OR avia*) AND trend*) which resulted in 313 references.
4. ALL=(birds AND species richness AND temporal trend) which resulted in 88 references.

For each query, the title and abstract of these articles were reviewed. In addition, we scanned the references of these articles for other potentially relevant literature. When the temporal trend was explicitly reported (either in a graph or text), we extracted from the material and methods the type of metric, the spatial grain of the trend (i.e. the area at which the metric trend is assessed), its temporal grain, the spatial extent (i.e. the entire area on which the study applies), the temporal extent and the beginning and ending years of the study as well as the general trend of the metric (Table 2.1). We discretized spatial grain sizes discretized into four levels: local $\leq 25 \text{ Km}^2$, regional $> 25 \text{ Km}^2$, national when an entire country was considered, and global at the worldwide scale (grain = extent = the entire Earth’s mainland).

Concerning the trend assessment, different papers contain the p-value or directly specify the significance of a trend of a metric. However, some papers give only graphical representations of the trend. For those,

the standard error was used when displayed. If a study only gave the trend, this is noted in the column Note of the Table 6.1. Moreover, the final trend retained (i.e. either Increase, Stable or Decrease) doesn't reflect all the fluctuations of the metric through time but rather the difference between the starting and ending points.

We found 31 references in which authors were both determining the temporal trend of a metric and explicitly defining the grain size. However, only 17 of them used spatial replicates and were thus relevant for this study (Table 3.1).

The classes of metric are: *Species richness*, *Evenness*, *Abundance*, *Diversity*, *Temporal beta-diversity*, *Spatial beta-diversity*, *Functional diversity*, *Functional evenness*, *Functional richness*. Some of these classes contain several different indexes. This is the case for instance for the class *Diversity*, which can be defined by either the Shannon or the Simpson index, or for the class *Abundance* which contains various multi-species indicators (for more information, see Table 9.1).

Table 3.1: Trends of different metrics of biodiversity at various spatial and temporal scales

Reference	Metric	Spatial grain (Km ²)	Temporal grain (year)	Spatial extent (Km ²)	Temporal extent (year)	Years	Country	Trend
Barnagaud et al., 2017	Evenness	Local	1.0	9834000	41	1970-2011	USA	Increase
	SR	Local	1.0	9834000	41	1970-2011	USA	Increase
Bowler et al., 2021	Abundance	National	1.0	520475	27	1990-2016	Czech Rep., Switzerland, Denmark, Germany	Stable
Chase et al., 2019	SR	Local	5.0	2800000	30	1982–2011	USA, Canada	Stable
	SR	Regional	5.0	2800000	30	1982–2011	USA, Canada	Stable
	SR	Regional	5.0	2800000	30	1982–2011	USA, Canada	Stable
	SR	Local	5.0	2800000	30	1982–2011	USA, Canada	Increase
	SR	Local	5.0	2800000	30	1982–2011	USA, Canada	Increase
Chiron et al., 2013	Abundance	Regional	1.0	643801	14	2007-2020	France	Decrease
	Abundance	Regional	1.0	643801	14	2007-2020	France	Decrease
	Abundance	Regional	1.0	643801	14	2007-2020	France	Decrease
Davey et al., 2012	Diversity	Local	1.0	242495	13	1994-2006	UK	Increase
	Evenness	Local	1.0	242495	13	1994-2006	UK	Increase
	SR	Local	1.0	242495	13	1994-2006	UK	Increase

Table 3.1: Trends of different metrics of biodiversity at various spatial and temporal scales (*continued*)

Reference	Metric	Spatial grain (Km ²)	Temporal grain (year)	Spatial extent (Km ²)	Temporal extent (year)	Years	Country	Trend
Harrison et al., 2014	Abundance	Local	1.0	200000	18	1994-2011	Great Britain, UK	Increase
	Abundance	Local	1.0	200000	18	1994-2011	Great Britain, UK	Stable
	Abundance	Local	1.0	200000	18	1994-2011	Great Britain, UK	Stable
Harrison et al., 2016	Abundance	Local	0.5	242495	20	1994-2013	UK	Increase
	Abundance	Local	0.5	242495	20	1994-2013	UK	Stable
Jarzyna and Jetz, 2018	Abundance	Local	0.5	242495	20	1994-2013	UK	Stable
	SR	Regional	1.0	9834000	45	1969-2013	USA	Increase
	SR	Regional	1.0	9834000	45	1969-2013	USA	Increase
	SR	Regional	1.0	9834000	45	1969-2013	USA	Increase
	SR	National	1.0	9834000	45	1969-2013	USA	Increase
	SR	Global	1.0	148940000	45	1969-2013	World	Decrease
	Temporal beta-diversity	Regional	1.0	9834000	45	1969-2013	USA	Increase
	Temporal beta-diversity	Regional	1.0	9834000	45	1969-2013	USA	Increase
	Temporal beta-diversity	Regional	1.0	9834000	45	1969-2013	USA	Increase

Table 3.1: Trends of different metrics of biodiversity at various spatial and temporal scales (*continued*)

Reference	Metric	Spatial grain (Km ²)	Temporal grain (year)	Spatial extent (Km ²)	Temporal extent (year)	Years	Country	Trend
	Temporal beta-diversity	National	1.0	9834000	45	1969-2013	USA	Increase
	Temporal beta-diversity	Global	1.0	148940000	45	1969-2013	World	Stable
Pilotto et al., 2020	Abundance	Local	NA	10180000	NA	NA	Europe	Stable
	Diversity	Local	NA	10180000	NA	NA	Europe	Increase
	SR	Local	NA	10180000	NA	NA	Europe	Increase
	Temporal beta-diversity	Local	NA	10180000	NA	NA	Europe	Stable
Ram et al., 2017	Abundance	Local	1.0	350000	18	1998-2015	Sweden	Increase
	SR	Regional	1.0	350000	18	1998-2015	Sweden	Increase
Reif et al., 2013	Spatial beta-diversity	Local	1.0	79000	23	1982-2004	Czech Rep.	Stable
	SR	Local	1.0	79000	23	1982-2004	Czech Rep.	Stable
	SR	National	1.0	79000	23	1982-2004	Czech Rep.	Stable
Schipper et al., 2016	Abundance	Local	5.0	24710000	40	1971-2010	Canada, USA, Mexico	Increase
	Diversity	Local	5.0	24710000	40	1971-2010	Canada, USA, Mexico	Increase
	Diversity	Local	5.0	24710000	40	1971-2010	Canada, USA, Mexico	Increase

Table 3.1: Trends of different metrics of biodiversity at various spatial and temporal scales (*continued*)

Reference	Metric	Spatial grain (Km ²)	Temporal grain (year)	Spatial extent (Km ²)	Temporal extent (year)	Years	Country	Trend
Sorte and Boecklen, 2005	Functional diversity	Local	5.0	24710000	40	1971-2010	Canada, USA, Mexico	Decrease
	Functional evenness	Local	5.0	24710000	40	1971-2010	Canada, USA, Mexico	Increase
	Functional richness	Local	5.0	24710000	40	1971-2010	Canada, USA, Mexico	Increase
	SR	Local	5.0	24710000	40	1971-2010	Canada, USA, Mexico	Increase
	Abundance	Local	1.0	9834000	36	1968-2003	USA	Decrease
	Evenness	Local	1.0	9834000	36	-	USA	Decrease
	SR	Local	1.0	9834000	36	1968-2003	USA	Increase
	SR	Regional	4.0	41543	28	1973-2000	Netherlands	Increase
	SR	Local	4.0	41543	28	1973-2000	Netherlands	Increase
	SR	National	4.0	41543	28	1973-2000	Netherlands	Increase
Wretenberg et al., 2010	SR	Local	1.0	1800	11	1994-2004	Sweden	Decrease
Inger et al., 2015	Abundance	National	1.0	10180000	22	1980-2001	Europe	Decrease
Donald et al., 2001	Abundance	National	NA	10180000	21	1970-1990	Europe	Decrease

4. Spatial scale and temporal trends

Overall trends. The median spatial extent of the 17 articles is $5.20475 \times 10^5 \text{ Km}^2$, with the smallest area of 1800 Km^2 and the greatest representing the global emerged surface (*i.e.* $1.4894 \times 10^8 \text{ Km}^2$). These articles reported XXX combinations of scales and metrics. Overall, there were 11 *Decrease*, 31 *Increase* and 14 *Stable* reliable trends (*i.e.* spatially replicated) across the literature. In our case, local scales are more represented than the others and the number of articles decreases with the increasing spatial scale (Figure 3.1). This is expected, as the spatial replications get more demanding in organization and resources as the grain size enlarges. The Increase of the metrics seems to be dominating at smaller scales. On the other hand, the proportion of Decrease is larger at regional scales than at local scales. At the global scale, no Increase was found.

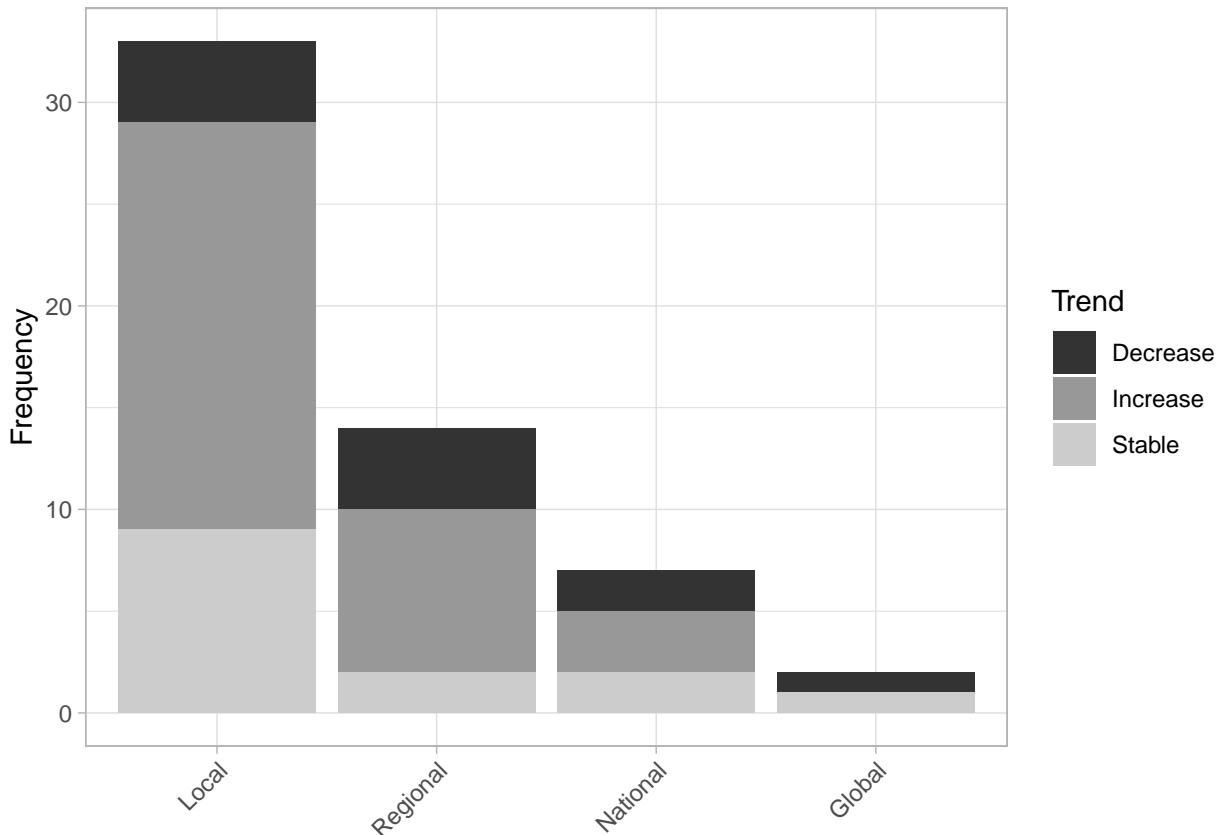


Figure 4.1: Proportion of *Increase*, *Decrease* or *Stable* trends for each spatial scale

Trends by metric. Among the different metrics, most of the examined studies deal with temporal trends of species richness and abundance (Figure 3.2). The less common trend of abundance is Increase, whilst Decrease and Stability are both as common. Diversity indexes (*i.e.* Sørensen and Jaccard) were always

found increasing and temporal β -diversity was found most of the time increasing and never decreasing. The use of the other metrics was rare.

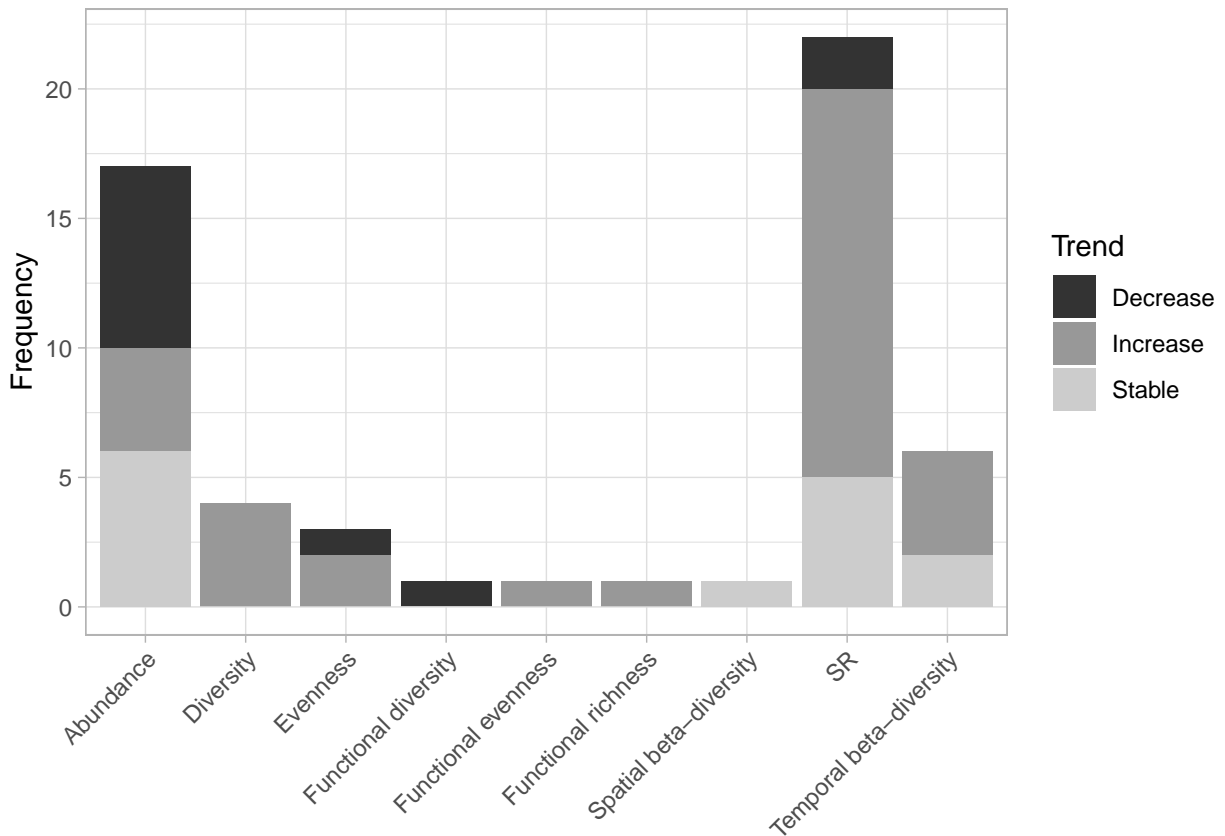


Figure 4.2: Proportion of *Increase*, *Decrease* or *Stable* trends for each of the metric

Trends by spatial grain. In the studies that we reviewed, at local grains, species richness mostly increased (Figure 3.3). Evenness indices, i.e. taxonomic and functional evenness, were also mostly increasing. Concerning the abundance indices, we found mostly no trend, or increases. At regional grains, abundance metrics always decreased, temporal β -diversity always increased, and species richness increased. At national and global grains, studies analyzing multiple locations are rare, and most trends reported here for these two spatial scales are not replicated. Exceptions are D. Bowler et al. (2021) and Donald, Green, and Heath (2001). The former showed negative trends in abundance indices for Denmark and Germany, and positive trends for Switzerland and Czech Republic, i.e. no clear direction of the trend (here referred as Stable). However, for Donald, Green, and Heath (2001), trends of mean population size were computed for 30 European countries and the majority was negative. Global trends only come from Jarzyna and Jetz (2018), who report a global decrease of species richness and a stable temporal beta-diversity.

However, concerning species richness, we can see that the decrease is global, but that this decrease is

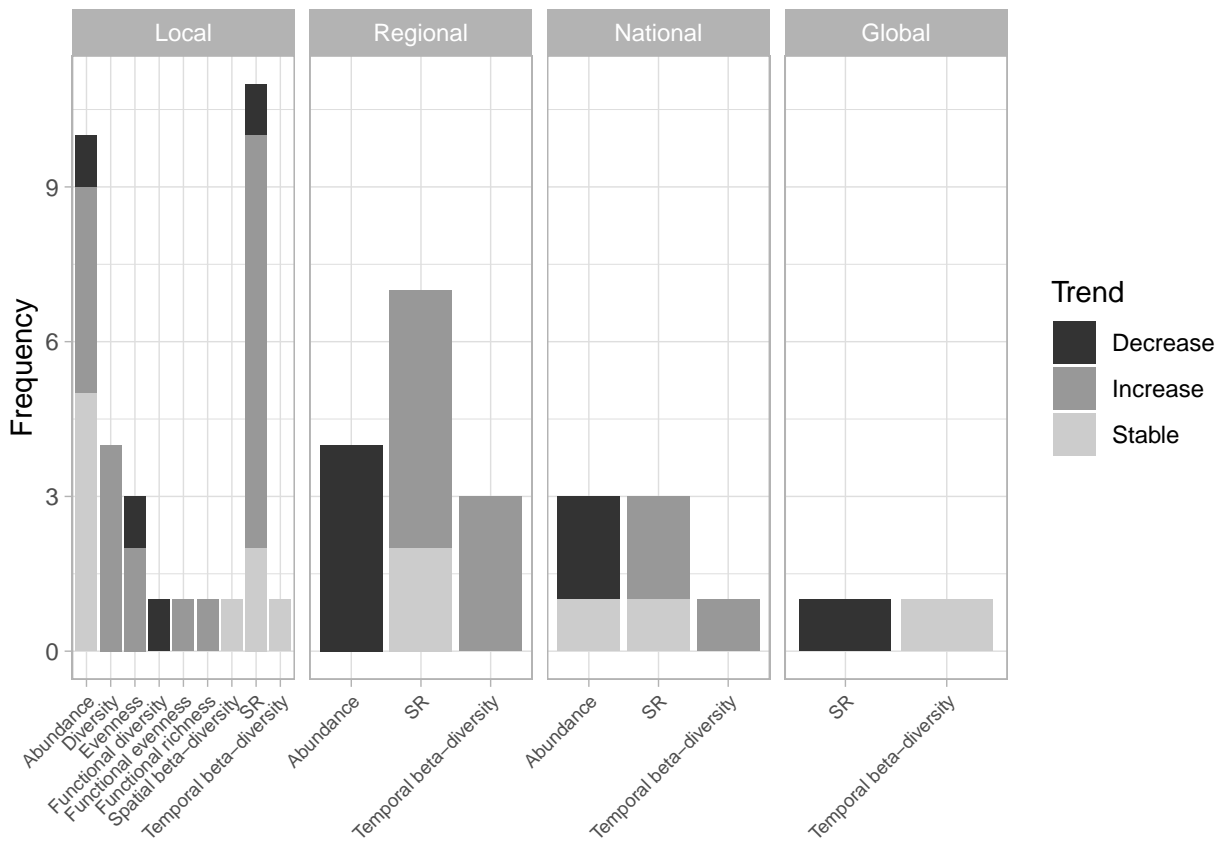


Figure 4.3: Proportion of *Increase*, *Decrease* or *Stable* trends for each metric. Each panel represent one spatial scale

rare at finer spatial scales. In fact, we observe more often increases, confirming the high perturbations that biodiversity is undergoing (Dornelas et al. 2014; Vaidyanathan 2021). This analysis goes along with the temporal β -diversity which is always observed either stable or increasing. Indeed, increasing turnover through time is a sign of an increasing perturbation of the ecosystems.

5. Temporal scale and temporal trends

The oldest study started in 1968 and the median duration is 23 years, with a minimum time-span of 11 years and a maximum of 45. We found longest temporal grains in studies with large spatial extent (Figure 4.1). This is because data used in the selected article are mainly structured data, i.e. data following a well established sampling plan. This type of survey is sparse since it needs resources and organization. Increasing the spatial extent thus increases both the temporal extent and the temporal grain. This explains this positive correlation between spatial extent and temporal grain. This limitation can be overcome thanks to citizen science data, which have increasingly been used (e.g. D. E. Bowler et al. 2021; Isaac et al. 2020, 2014). The opportunistic nature of these data allows for short census times, even over a large area. These data, with high temporal grain resolution for large spatial scales, could in future be used to explore in more details the temporal scaling of biodiversity trends.

Determination of the temporal grain in the studies in Table 1 was complicated. Usually, the temporal grain of the sampling was specified, but sometimes with inaccuracies (e.g. “in the early morning”). Moreover, the temporal grain of the sampling doesn’t represent the temporal grain of the metric. For instance, some metrics are summed over a certain area (e.g. summing the species richness over an atlas square, such as in Van Turnhout et al. 2007). Analogically, the temporal grain should have also been summed over all the sampling censuses englobed in this area, but this was never the case. Moreover, when the trend is computed, usually the lag (i.e. the time between two computation of the metric, Dungan et al 2002) is clearly specified, but the temporal grain is not.

For the cases where the metric of biodiversity is determined out of model (e.g. Harrison et al. 2014), it is easier to assess the temporal grain, since predictions allow one to extrapolate the data from the sampling temporal grain to a wanted one. Thus, for these cases, the final temporal grain was most of the time explicitly given.

In short, temporal grain is rarely specified, especially when the metric computed makes it different from

A scatter plot showing the relationship between Temporal grain (year) on the y-axis and Spatial extent (Km²) on the x-axis. The y-axis ranges from 1 to 5, and the x-axis is on a logarithmic scale from 1e+04 to 1e+08. There are 15 data points represented by black dots. The data points are as follows:

Spatial extent (Km²)	Temporal grain (year)
~3,000	1
~40,000	4
~80,000	1
~200,000	1
~250,000	1
~300,000	0.5
~400,000	1
~500,000	1
~600,000	1
~800,000	1
~3,000,000	5
~10,000,000	1
~30,000,000	5
~150,000,000	1

6. Lack of spatial replication

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Another common type of study uses the *space-for-time substitution* (Walker et al. 2010) to assess the trend of a metric (one of the best example is J. K. Hill and Hamer 2004). This method consists in considering sampling in different places as representing a temporal trend. One could think that using these studies could increase significantly the spatial replicates. However, the space-for-time substitution is mainly used to assess the impact of a processes (e.g. before/after logging, before/after urbanization etc) meaning that the trend computed is highly biased, which we try to avoid for our topic.

Even fewer articles computed the trends of metrics with spatial replicates across more than one spatial grain. So far this was the case for only Chase et al. (2019) and Jarzyna and Jetz (2018). Importantly, Jarzyna et al. (2015) did spatial replicates of temporal change community metrics (i.e. temporal dissimilarity, temporal turnover, extinction and colonization) at several spatial scales. However, the temporal trend of these metrics weren't considered and are therefore not reported in Table 2.1.

7. Metric heterogeneity

In contrast to macroecology, applied ecology has offered multiparametric indices that aim to reflect multiple components of an ecosystem, the so-called *composite indicators*. Examples are the red list index (Stuart H. M. Butchart et al. 2007) or biodiversity change indices (Normander et al. 2012). The latter have proven effective for conservation policies (REF). For birds, these indices have been widely used (see review by Fraixedas et al. 2020), while the use of simpler macroecological indices (e.g. species richness, beta diversity) is uncommon.

8. Future directions

A striking but expected (see Meyer et al. 2018), was the lack of studies from outside of the high-income global North. Out of 17 papers, 5 were located in North America and 12 in Europe. This gap was also reported by Fraixedas et al. (2020). Yet, biodiversity dynamic in Europe may not be representative of global dynamic, and studies of biodiversity trends at several spatio-temporal scales are needed outside of Europe. These studies are needed on local grains, as well as at the spatial grain of continents (e.g. as has been done by Alroy 2015 for amphibians and reptiles).

The spatial grain of biodiversity trends is critical, yet this has not always been specified in the articles. One should consider the way a metric was computed, e.g. if it was summed, modeled, averaged over

the sampling units. According to the method, the spatial grain can vary from the sampling unit to the sampling extent. Given the importance of spatial scaling of biodiversity patterns (Storch and Gaston 2004), one has to expect that it will be also important for its dynamic (e.g. Chase et al. 2019). We thus argue that authors should pay extra attention to specifying the spatial grain for every metric of biodiversity trends.

The importance of temporal scaling of biodiversity is known since Grinnell (1922), who used California birds to demonstrate the species-time relationship, which has since been proven to be common with other bird populations (White 2004). Thus, as spatial grain, temporal grain is known to be important to explicit. However, there is no consensus on the definition of temporal grain and is thus specified in various ways: sometimes very precised (e.g. time of each census point, as in Schipper et al. 2016) and sometimes without explicit information (e.g. Harrison et al. 2014: ‘The sites are visited twice a year (April to early May and late May to June), during which volunteers walk two parallel 1-km-long transect lines [...]’). As for the spatial grain, the metric can vary according to the temporal grain and the way it is computed. However, when temporal trend of a metric is assessed, the temporal lag of the trend is often only specified (i.e. time-span between t and $t+1$). In other words, when computing a trend, one usually uses a single point either every day, every month or every year. However, this temporal lag doesn’t represent the temporal grain that interests us. If one wants to study the temporal scaling of biodiversity trends, a clear assessment of the temporal grain needs to be done systematically.

9. Conclusion

Reviewing the scientific literature on avian biodiversity trends give us a glimpse of what needs to be done to better understand the scaling of biodiversity dynamic. The first challenge is to find a common definition of spatial and temporal grain when computing the trend. We showed that these definitions vary according to the way the metric is computed. Whilst spatial grain of a trend is intuitively important for biodiversity, temporal grain is less often considered, especially when the temporal grain is computed. Finally, as birds are one of the most data-rich taxa of vertebrates, the challenges highlighted here will be even more severe for other groups.

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Supplementary materials

Table 9.1: Supplementary informations about each article

Reference	Spatial grain (Km ²)	Trend	Note
Barnagaud et al., 2017	Local	Increase	Not sure that it is at the road scale: "Taxonomic evenness showed a marginal, yet significant, non-linear increase from close to 0.54 in the first decade to 0.56 in the last decade (Table 1), suggesting a light trend towards a more even distribution of species' abundances among species within local assemblages "
	Local	Increase	Mean change of SR at the road scales Area of the road = $(40/0.8) * (\pi * 4002)$ with a road of 40 Km with point counts spaced by 0.8 Km and a census radius of 400m
Bowler et al., 2021	National	Stable	Metric = MSI, as many and as intense increase (i.e. Czech Rep. and Switzerland) than decrease (i.e. Germany and Denmark)
Chase et al., 2019	Local	Stable	NA
	Regional	Stable	NA
	Regional	Stable	NA
	Local	Increase	NA
	Local	Increase	NA
Chiron et al., 2013	Regional	Decrease	Concerning the spatial scale, predictions are made using the spatial unit of 4 Km ² and the FBI is computed for each region of France, then meanned. Prediction with baseline scenario
	Regional	Decrease	FBI prediction with CAP greening cenario
	Regional	Decrease	FBI prediction with No Pillar I scenario
	Regional	Decrease	FBI prediction with biofuel scenario
Davey et al., 2012	Local	Increase	Metric = Simpson.They predict the metric using a GAM with spatial resolution of 1 Km ² . Then they show the trend for the mean value of the metric per year
	Local	Increase	NA
	Local	Increase	NA

Table 9.1: Supplementary informations about each article (*continued*)

Reference	Spatial grain (Km ²)	Trend	Note
Harrison et al., 2014	Local	Increase	To assess the metric, they use a GAM to predict the abundance over the entire area of interest (spatial resolution = 1 Km ²) and then compute the geometric mean of species abundance = Multi Species Index (as in Studeny et al., 2013) from the prediction. Data used to learn the GAM are sampled from plots of 1 Km ² . Farmland communities
	Local	Stable	Farmland communities, GoF ($\lambda = -1$) = weighted towards the rare species
	Local	Stable	Farmland communities, GoF ($\lambda = -2$) weighted towards the common species
Harrison et al., 2016	Local	Increase	Geometric mean of species abundance, they predict the abundance with resolution of 1 Km ² and then computed the metric for each 10000 Km ² cell across Great Britain, Visited twice a year
	Local	Stable	GoF ($\lambda = -1$) = toward rare species" The goodness-of-fit-based measure of biodiversity suggests that both rare and common species made gains through much of Britain in the first half of the time period, and losses in the second half.", Visited twice a year / Increase first half and second second halfGoF ($\lambda = -1$)
	Local	Stable	GoF ($\lambda = -2$) = toward common species " The goodness-of-fit-based measure of biodiversity suggests that both rare and common species made gains through much of Britain in the first half of the time period, and losses in the second half.", Visited twice a year / Increase first half and second second half
Jarzyna and Jetz, 2018	Regional	Increase	NA
	Regional	Increase	NA
	Regional	Increase	NA
	National	Increase	NA
	Global	Decrease	NA
	Regional	Increase	NA
	Regional	Increase	NA

Table 9.1: Supplementary informations about each article (*continued*)

Reference	Spatial grain (Km ²)	Trend	Note
	Regional	Increase	NA
	National	Increase	NA
Pilotto et al., 2020	Global	Stable	NA
	Local	Stable	"Analyses of the trends in local biodiversity over large spatial scales"
	Local	Increase	Metric = Simpson, "Analyses of the trends in local biodiversity over large spatial scales"
	Local	Increase	"Analyses of the trends in local biodiversity over large spatial scales"
	Local	Stable	"Analyses of the trends in local biodiversity over large spatial scales"
Ram et al., 2017	Local	Increase	MSI for forest species, road of 8 Km with no limitations so assumed 200m
	Regional	Increase	SR for forest species meaned over roads, spatial grain = 8* .4 with road of 8 Km and census radius no limitations so assumed 200m
Reif et al., 2013	Local	Stable	Jaccard index, pairwise comparisons between transects
	Local	Stable	JPSP data, transect scale
	National	Stable	JPSP data, national scale
Schipper et al., 2016	Local	Increase	The metric (i.e. geometric mean) is meaned over each road. Area of the road = 50*(pi*4002) with 50 census point per road and a census radius of 400m
	Local	Increase	Metric = Shannon
	Local	Increase	Metric = Simpson
	Local	Decrease	NA
	Local	Increase	NA
	Local	Increase	NA
	Local	Increase	NA

Table 9.1: Supplementary informations about each article (*continued*)

Reference	Spatial grain (Km ²)	Trend	Note
Sorte and Boecklen, 2005	Local	Decrease	NA
	Local	Decrease	Metric = evenness
	Local	Increase	The metric is meaned over each road. Area of the road = $50 \cdot (\pi \cdot 400^2)$ with 50 census point per road and a census radius of 400m
Van Turnhout et al., 2007	Regional	Increase	For each region, the trend is computed using the mean number of species per atlas square
	Local	Increase	Mainly increase of SR but the proportion of negative trend were higher than for the regional scale
	National	Increase	National scale
Wretenberg et al., 2010	Local	Decrease	looking at the trend through different environmental policies, " local species richness (i.e. at the scale of sites = 3 hectares) decreased significantly probably as a result of an overall reduced abundance of several species. "
Inger et al., 2015	National	Decrease	NA
Donald et al., 2001	National	Decrease	The metric is referred as "Mean population" and the trend is estimated for each european country