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3 **Are global forests performing in sync? The need to account for spatiotemporal**  
4 **biases in tree-ring records**

5

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22

23 **Abstract**

24 Populations whose dynamics are driven by correlated environmental stochasticity face  
25 increased risk of extinction. Forests in particular are being pushed to their physiological  
26 limits under global change; hence, the analysis of common patterns of tree performance  
27 across scales becomes crucial to discern early warning signals and tipping points. Here,  
28 we critically evaluate customary and recent approaches for the analysis of time series of  
29 radial growth based on simple correlations as measure of direction and signal strength  
30 shared by spatially-segregated forests (synchrony). By accounting for changes in spatial  
31 distribution of populations and temporal coverage of tree-ring records, we show that  
32 growth synchrony has not substantially augmented for the past 150 years across Europe,  
33 a continent with high availability of tree-ring series. We guard against the use of absolute  
34 correlations as a metric of synchrony and stress that robust analytical methods should be  
35 applied to evaluate synchrony trends when using biased spatiotemporal databases.

36

37 **Significance Statement**

38 The increasing availability of tree-ring records globally constitutes an invaluable resource  
39 to spot changes in forest dynamics at varying spatiotemporal scales. In this context,  
40 assessments of changes in tree growth synchrony are useful to anticipate the fate of forest  
41 ecosystems in a warmer world. While relevant patterns may emerge from global  
42 syntheses, we are concerned about approaches using absolute correlations as synchrony  
43 metrics which may mistakenly lead to the conclusion that trees are progressively  
44 performing in sync. We encourage more systematic evaluations of synchrony trends at  
45 meaningful biogeographical scales as effective approach to interpret forest dynamics  
46 under global change.

47

48 **Keywords:** biogeographical patterns, climate change, dendroecology, International Tree-  
49 Ring Data Bank, mixed models, Moran effect, growth synchrony

50

## 51 **1. The geography of spatial synchrony: the case of forest growth dynamics**

52 In ecology, spatial synchrony refers to parallel changes in time-varying features  
53 of geographically disjunct populations (Liebhold, Koenig, & Bjørnstad, 2004). The main  
54 factor engendering population synchronization is assumed to be climatic forcing (i.e., the  
55 Moran effect; Moran, 1953). Over the past years, increasing attention has been given to  
56 the study of the temporal coherence of tree radial growth in the Anthropocene across  
57 bioclimatic gradients (e.g. Briffa et al., 2008; Shestakova et al., 2016, 2018, 2019a,b;  
58 Ponocná et al., 2018; Camarero et al., 2021). As a result, intricate and non-generalized  
59 temporal trends of common variability of tree growth have been documented, which point  
60 to biophysical complexities of the role of emerging abiotic (climatic) limitations and  
61 related factors (outbreaks of pests and diseases, fires, etc.) impacting on tree performance,  
62 especially since mid-20<sup>th</sup> century. A recent examination of ring-width records  
63 (Manzanedo, Hille Ris Lambers, Rademacher, & Pederson, 2020) concluded that spatial  
64 synchrony of tree growth strongly (and consistently) rose worldwide in the past few years.  
65 This outcome stemmed from the global analysis of tree-ring chronologies available at the  
66 International Tree-Ring Data Bank (ITRDB). Oddly, this phenomenon went unnoticed  
67 based on evidence collected so far at local and regional scales, where either positive,  
68 stable or negative trends had been reported (Briffa et al., 2008; Shestakova et al., 2016,  
69 2018, 2019a,b; Ponocná et al., 2018; Camarero et al., 2021).

70 The present correspondence and associated analysis were originally motivated by  
71 such a report, which in any case was greatly welcomed owing to its potential for providing  
72 a foundation towards understanding the global performance of disjunct forests and its

73 implications for the terrestrial carbon sink. Despite the fact that the paper by Manzanedo  
74 et al. (2020) was recently retracted due to “a coding error, correction of which undermines  
75 the main conclusions of the study (...)” (Manzanedo, Hille Ris Lambers, Rademacher, &  
76 Pederson, 2021), we believe the proposed methodology to infer growth synchrony trends  
77 – based on the analysis of absolute correlations – deserves an additional warning, merely  
78 because in ecological applications both the sign and strength of the simple correlation  
79 coefficient matter. Thus, our (re)evaluation may still provide relevant insights to the  
80 dendro- and, in general, the ecological community while serving as cautionary tale about  
81 how inappropriate methodologies can yield results that are purely artefactual. In turn, we  
82 would also like to draw attention to the prevailing spatiotemporal biases in the ITRDB  
83 database, in particular to the decrease in availability and length of updated tree-ring series,  
84 which warrant careful examination through the application of suitable analytical  
85 approaches.

86

## 87 **2. The analysis of trends in growth synchrony: methodological considerations**

88 Traditional methods for assessing synchrony patterns in ring-width chronologies  
89 (or other tree-ring traits) include parametric and non-parametric approaches. Among the  
90 first, simple (averaged pair-wise) and intra-class (variance-based) correlations provide  
91 straightforward metrics for characterizing the strength and nature of the signal shared  
92 among time series (Shestakova et al., 2018). The latter, grounded in mixed modelling  
93 principles, is also well suited to unbalanced datasets (Shestakova et al., 2016). Non-  
94 parametric choices are, among others, the “Gleichläufigkeit” coefficient of coincidence  
95 (Eckstein & Bauch, 1969) and Kendall’s concordance coefficient (Kendall, 1975). The  
96 latter is less sensitive to deviations from normality (as ring-width indices tend towards

97 asymmetry), while it is claimed to detect similarities in variability between series better  
98 than simple correlations (Briffa et al., 2008).

99         Alternatively, the mean absolute correlation between pairs of chronologies has  
100 been recently proposed as a metric of synchrony (Manzanedo et al., 2020). Although this  
101 methodology might be suitable to unveil increasingly large (both positive and negative)  
102 coherent growth responses being ‘in sync’ at large distances due to climatic forcing (i.e.  
103 climate dipoles), it has two relevant problems. First, it mistreats opposing values of the  
104 simple (pair-wise) correlation coefficient, hence resulting in a (fully predictable) positive  
105 synchrony between tree-ring chronologies – even when there is none. Indeed, ignoring  
106 the sign of the correlation coefficient will inevitably fallout in artificially inflated positive  
107 relationships. Second, overlooking the nature (sign) and geographical structure of the  
108 common signal shared by a set of chronologies hampers the mechanistic interpretation of  
109 large-scale drivers of tree growth, thereby obstructing the identification of on-going  
110 ecological repercussions of global change across regions and biomes. It must be noted  
111 that the presence of sizeable negative correlations at large distances is highly unlikely  
112 (Shestakova et al., 2018), as it is at odds with the observation that atmospheric circulation  
113 patterns often affect neighbour regions very disparately (Allen, Breshears, & McDowell,  
114 2015). As an alternative to absolute correlations, correlograms are an obvious choice to  
115 evaluate the spatial distribution of common growth patterns among pairs of chronologies  
116 (Bjørnstad, Ims, & Lambin, 1999; Koenig, 1999). They are grounded in simple  
117 correlations and facilitate understanding and testing the magnitude and nature of the  
118 metric across scales.

119         An additional matter of concern related to the use of absolute correlations as a  
120 metric of synchrony is the sharp decrease in availability of ITRDB series occurring  
121 immediately before the turn of this century. This bias is widely acknowledged by the

122 dendro-community (Babst, Poulter, Bodesheim, Mahecha, & Frank, 2017; Zhao et al.,  
123 2019). Indeed, a change in geographic coverage of chronologies coupled with a poorer  
124 temporal coverage in recent years can be responsible of a sudden increase in mean  
125 absolute correlation simply because of higher chances of (large) spurious correlations  
126 contributing to this metric.

127

### 128 **3. An example: Europe-wide analysis of tree growth across the Anthropocene**

129 In order to critically evaluate the performance of the aforementioned  
130 methodologies, which are based on correlation principles of temporal signals, we (i)  
131 carried out a reanalysis of European tree-ring chronologies (990 chronologies; 24.5% of  
132 the ITRDB database) following the methodology applied in Manzanedo et al. (2020), and  
133 (ii) compared the results with alternative approaches for evaluating synchrony trends in  
134 tree growth (i.e. averaged pairwise and intra-class correlations; Shestakova et al., 2018).  
135 The metrics were calculated for 50-year periods to retrieve moving-window estimates of  
136 synchrony trends. The rise in synchrony estimated from absolute correlations mostly  
137 vanished once the dataset was filtered for chronologies spanning at least 25 years within  
138 each 50-year period (Figure 1a). In fact, the spatial distribution of pairwise correlations  
139 for equally spaced 50-year periods, starting in 1850, showed an increase of spurious  
140 associations achieving either high or low erratic values during the period of 1967–2016  
141 (Supplementary Figure S1). This observation coincided with an acute decrease in  
142 chronology length (Figure 1b), distance between sites and number of available  
143 chronologies (Supplementary Figure S2). If raw (not absolute) correlations were used,  
144 the rising trend in synchrony disappeared (Figure 1a). By using the intra-class correlation  
145 metric of synchrony (Shestakova et al., 2018), which efficiently deals with the unbalance  
146 of partially overlapping tree-ring data (but not with changes in their spatial coverage), the

147 increasing trend in synchrony could still be observed, but in a much lower degree (Figure  
148 1a). Indeed, these results put into question the existence of a global trend towards rising  
149 temporal coherence of tree growth. Although we restricted our analyses to Europe, a  
150 continent with the highest availability of tree-ring series (together with North America),  
151 other regions are likely to display similar patterns.

152

#### 153 **4. Interpretation and attribution of changes in spatial synchrony of forest growth**

154 The attribution of synchronous tree growth dynamics at large spatiotemporal  
155 scales to exogenous (e.g. climate) factors is highly relevant to forecast potential  
156 ecological consequences of global change and the role of forests as carbon sinks.  
157 Unquestionably, attribution analyses should be carefully performed and interpreted.  
158 However, bigger efforts have been customarily placed on ascertaining relationships  
159 between tree rings and climate for stability at local rather than broader (regional,  
160 continental) scales (Allen et al., 2018). In this regard, disentangling the role of long- and  
161 short-term changes in exogenous variables on the geographical patterns of synchrony in  
162 tree-ring networks can be problematic. This is mainly because of two independent issues:  
163 an imperfect availability of spatially-explicit environmental data with sufficient  
164 resolution, and the spurious regression problem that arises when linking unrelated non-  
165 stationary variables in longitudinal studies (Shestakova et al., 2019a).

166 Spurious regression refers to the case where a pair of independent, non-stationary  
167 series exhibiting similar trends shows too frequently a significant relationship (high  $r^2$   
168 and significant  $t$ -values) according to standard inference in ordinary least squares  
169 regression (i.e. nonsense regression; Murray, 1994). For example, a generalized rise in  
170 growth synchrony can be easily mistaken as an early sign of current warming and extreme  
171 weather events affecting forest ecosystems. This is relevant because sharp spikes in

172 synchrony may anticipate widespread swings in carbon sink strength as trees sync with  
173 global temperature. The same problem is applicable to alternative drivers of tree growth  
174 showing a generalized increment over the last decades, such as atmospheric CO<sub>2</sub>  
175 concentrations. On the other hand, our re-analysis (as shown in Figure 1) suggests that a  
176 global increase in temporal coherence of tree growth is unlikely to occur, at least until  
177 recently, owing to the (barely) existing shifts in synchrony observed across the entire  
178 European continent. In fact, if warmer temperatures and rising atmospheric CO<sub>2</sub> were to  
179 improve growing conditions, a decrease in synchrony could be expected as climate  
180 decreases its importance as a global “synchronizer” whilst local conditions (topography,  
181 soil, tree-to-tree interactions) gain relevance. The opposite interpretation – namely that  
182 warmer conditions progressively lead to both enhanced and more homogeneous growth  
183 – is therefore not tenable. Alternatively, increased warming-induced drought stress may  
184 make trees gradually limited by water shortage and increase synchrony at large (e.g.  
185 subcontinental) scales (Shestakova et al., 2016; Babst et al., 2019).

186         It is therefore obvious that searching for evidences of the role of temperature as a  
187 synchronizing cue controlling forest dynamics must properly address the spurious  
188 regression problem that plagues non-stationary time series (i.e. with variable mean and  
189 variance) analysis. Indeed, the role of potential (external) drivers of growth synchrony  
190 changes should be always examined at the high-frequency domain by accounting for  
191 stochastic non-stationary fluctuations. This can be readily achieved through e.g.  
192 cointegration analysis, which looks for stationary linear combinations of non-stationary  
193 random variables (Engle & Granger, 1987; Shestakova et al., 2019a). Importantly, it is  
194 highly possible for a pair of time series (e.g. synchrony changes and climate) to have  
195 weak/strong correlation but strong/weak cointegrating relationships (Murray, 1994).

196 In the end, by properly identifying local and regional environmental controls of  
197 forest ecosystems, spatial and species-specific growth patterns can be better connected  
198 with temperature and moisture variations within particular climatic zones. This should  
199 enable ecologically meaningful interpretations of synchrony changes for their respective  
200 forests (Babst et al., 2013; Shestakova et al., 2019b). Other relevant factors to tree  
201 performance such as global dimming, permafrost melting or N deposition may also play  
202 a role as drivers of synchrony depending on the spatiotemporal coverage of the analysis  
203 and the targeted ecosystem(s). In addition to the ultimate influence of such exogenous  
204 drivers, it is important to note that the ability of different forests to cope with short-term  
205 stresses and disturbances (e.g. an extreme drought) is related to phenotypic plasticity,  
206 which strongly varies among species or functional types. While some (plastic) species  
207 may better adapt to future conditions, others might undergo collapse or extinction  
208 (Nogués-Bravo et al., 2018).

209

## 210 **5. Concluding remarks**

211 Global change impacts on forests will likely be traced back with increasing  
212 precision and higher spatiotemporal coverage provided a larger number of  
213 dendrochronological records is progressively available worldwide (Babst et al., 2017).  
214 For this purpose, a particularly informative approach is the analysis and interpretation of  
215 synchrony patterns in comprehensive tree-ring networks. This effort can be pursued at  
216 different spatial scales depending on the desired research emphasis on local, regional or  
217 global factors controlling tree performance and its past and potential future effects. While  
218 relevant patterns may emerge from global syntheses, these should be carefully evaluated  
219 bearing in mind prevailing spatiotemporal biases in assembled datasets. Altogether, we  
220 strongly encourage systematic assessments of growth synchrony at meaningful

221 biogeographical scales as an effective approach to interpret forest dynamics under global  
222 change, and guard against the use of absolute correlations as a metric of synchrony,  
223 particularly in unbalanced datasets.

224

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229

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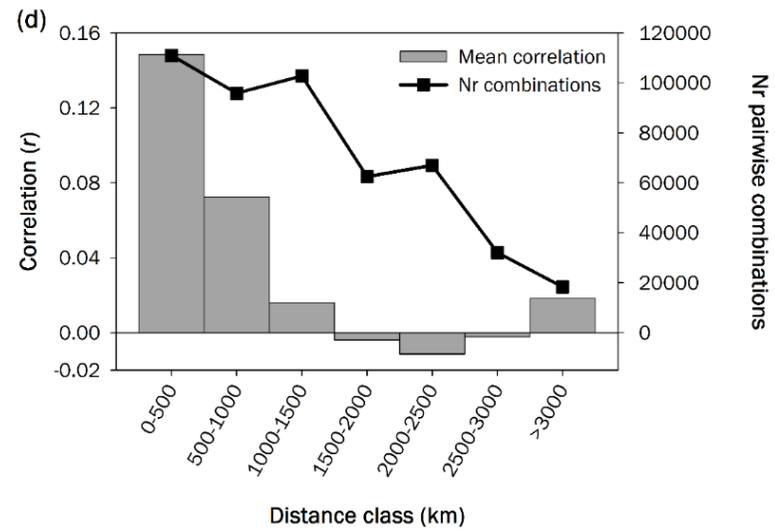
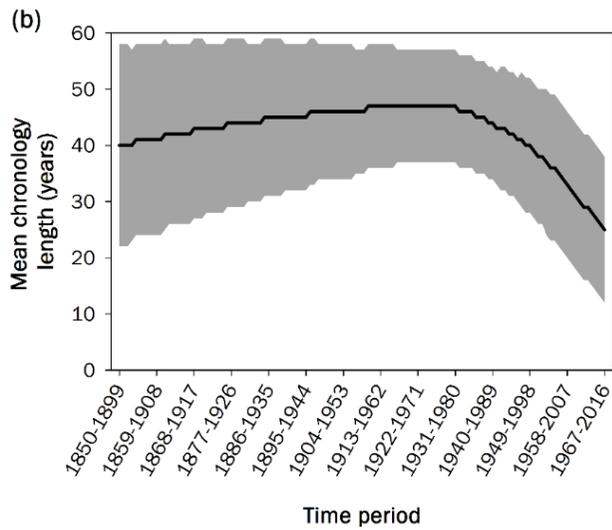
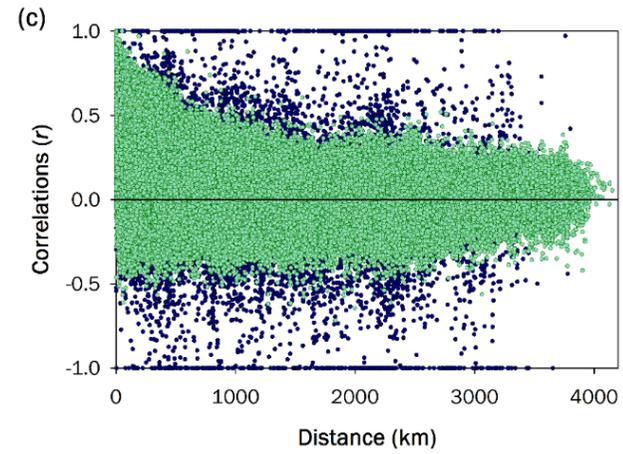
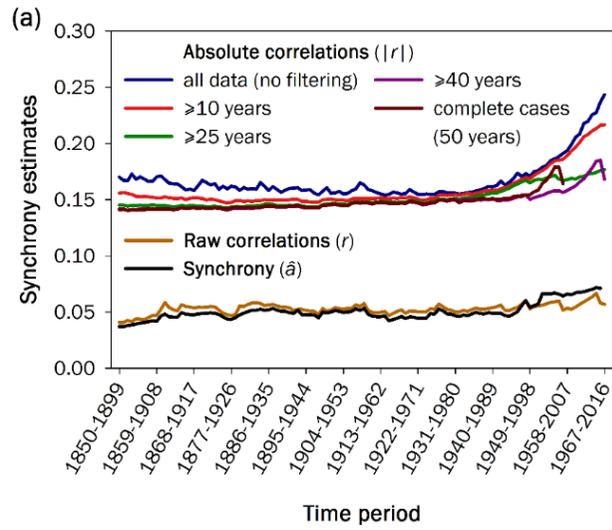
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**Figure 1.** Assessments of global growth synchrony are affected by spatiotemporal biases. **a**, Measures of tree growth synchrony across Europe based on ITRDB dataset calculated for 50-year periods lagged by 1 year over the period of 1850–2016 using absolute correlations ( $|r|$ ), raw correlations ( $r$ ) and intra-class correlations based on mixed modelling ( $\hat{a}$ ). Note that  $|r|$  values show a recent increasing trend driven by changes in the spatiotemporal coverage of the dataset, as suggested by **b**, a concomitant decrease in mean chronology length. A sensitivity analysis of  $|r|$  values filtering for increasingly high number of common (i.e. overlapping) years between pairs of chronologies (10, 25, 40 and 50 years) for each 50-year period is also presented, pointing to non-relevant temporal trends after controlling for chronology length. Note that  $|r|$  values for complete (i.e. 50-year length) chronologies are truncated starting from the period 1958–2007 (i.e. when the number of available chronologies at the ITRDB drops below 10% of the total amount). **c**, Pairwise correlations as a function of distance between (unfiltered) chronologies (blue dots) and **d**, correlogram of pairwise correlations showing a distance-dependent decay in mean correlation and a typical threshold of 1,000 km for sizeable common tree growth signals, respectively. Spurious (i.e. unusually high and low) correlations are evident across distances as the result of very limited temporal overlapping for particular pairs of chronologies, which disappear after filtering for pairs of chronologies including at least 25 common years (green dots in panel **c**). This problem gets worse for the most recent 50-year periods of the study period, hence boosting  $|r|$  (see Supplementary Figures S1 and S2).