

1      **Growth–reproduction trade-offs are common but**  
2      **changing in woody plants: a meta-analysis**

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4      Maciej K. Barczyk<sup>\*1</sup>, Michał Bogdziewicz<sup>1</sup>, Szymon M. Drobniak<sup>2,3</sup>, Maria Bogdańska<sup>1</sup>,  
5      Urszula Eichert<sup>1</sup>, Jessie J. Foest<sup>1</sup>, Valentin Journé<sup>1,4</sup>, Katarzyna Kondrat<sup>1</sup>, Jakub Szymkowiak<sup>1,5</sup>,  
6      Andrew Hacket-Pain<sup>6</sup>

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8  
9      <sup>1</sup> Forest Biology Center, Institute of Environmental Biology, Faculty of Biology, Adam Mickiewicz  
10     University, Uniwersytetu Poznańskiego 6, 61-614 Poznań, Poland

11     <sup>2</sup> Institute of Environmental Sciences, Faculty of Biology, Jagiellonian University, Krakow, Poland

12     <sup>3</sup> School of Biological, Environmental, and Earth Sciences, University of New South Wales, Sydney,  
13     Australia

14     <sup>4</sup> Department of Biology, Faculty of Science, Kyushu University, Fukuoka, Japan

15     <sup>5</sup> Population Ecology Research Unit, Institute of Environmental Biology, Faculty of Biology, Adam  
16     Mickiewicz University, Poznań, Poland

17     <sup>6</sup> Department of Geography and Planning, School of Environmental Sciences, University of Liverpool,  
18     Liverpool, UK

19  
20     \*Corresponding author: [maciejkbarczyk@gmail.com](mailto:maciejkbarczyk@gmail.com);

## 21 Abstract

22 Growth and reproduction draw on a common resource pool, yet empirical studies of woody plants  
23 report widely differing relationships between seed production and growth. Here we synthesize  
24 685 estimates from 78 studies covering 79 woody species to test how growth–reproduction corre-  
25 lations vary across time, species, and environments. Growth and reproduction measured within  
26 the same year were negatively correlated, suggesting an immediate cost of reproduction. Lagged  
27 growth–reproduction relationships further suggest that reproduction incurs delayed constraints  
28 on growth beyond the year of investment. The strength and direction of growth–reproduction  
29 correlations showed no detectable phylogenetic signal and were not systematically related to  
30 functional traits and climate. Instead, trade-offs were strongest in species with high interannual  
31 variability in seed production and weakened markedly over recent decades in these species.  
32 Together, these results show that growth–reproduction trade-offs in woody plants are common  
33 but not fixed, and that shifts in reproductive variability under environmental change can alter  
34 how trees balance growth and reproduction, with consequences for long-term forest functioning.

35 *keywords:* growth–reproduction trade-off | cost of reproduction | life-history theory |resource  
36 allocation | secondary growth | seed production | tree rings | masting | meta-analysis |

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## 38 Introduction

39 Individual fitness depends on survival, growth, and reproduction. These functions are con-  
40 strained by resource-based trade-offs, including the growth–mortality trade-off, growth–defense  
41 trade-off, and the cost of reproduction (Williams, 1966; Obeso, 2002; Dorken *et al.*, 2025;  
42 Russo *et al.*, 2020; Cople *et al.*, 2021). In long-lived organisms such as perennial plants,  
43 growth-reproduction trade-offs are expected because the allocation of resources is a zero-sum  
44 game; reproduction reduces resources available for other functions, such as growth and sur-  
45 vival, thereby lowering future fitness returns (Williams, 1966; Dorken *et al.*, 2025). Although  
46 the cost of reproduction in plants has been extensively discussed (Dorken *et al.*, 2025; Obeso,  
47 2002; Thomas, 2011), meta-analytic evidence is lacking on the extent to which woody plants  
48 face a compromise between seed production and somatic growth, and how this varies across  
49 species, traits, and habitats (Dorken *et al.*, 2025; Thomas, 2011). Systematic reviews suggest that  
50 growth–reproduction trade-offs are widespread (Obeso, 2002; Thomas, 2011), yet case studies  
51 report diverse outcomes, including negative (Norton & Kelly, 1988; Woodward *et al.*, 1994;  
52 Hadad *et al.*, 2021; Braun *et al.*, 2017), positive (Garcia-Barreda *et al.*, 2021; Alfaro-Sánchez  
53 *et al.*, 2015), and absent (Knops *et al.*, 2007; Patterson *et al.*, 2023) correlations between growth  
54 and reproduction.

55 Perennial plants vary widely in their annual allocation to reproduction, offering a test bed  
56 for growth–reproduction trade-offs (Norton & Kelly, 1988). Numerous long-lived woody plants  
57 show irregular seed production, with years of high, low or absent production, called masting  
58 or mast seeding (Bogdziewicz *et al.*, 2024). These among-year shifts in allocation create  
59 natural contrasts: years of heavy reproductive investment can be set against lean years to detect  
60 growth costs (Norton & Kelly, 1988; Monks *et al.*, 2016). The magnitude of these contrasts  
61 varies strongly among species, reflecting differences in interannual variability in reproductive  
62 investment, commonly quantified by the coefficient of variation of seed production ( $CV_p$ )  
63 (Pearse *et al.*, 2020; Journé *et al.*, 2023). Here, we synthesize multi-year measurements of  
64 seed production and growth to test how prevalent such trade-offs are across woody plants, and  
65 whether their strength depends on phylogeny, functional traits, climate, or temporal change  
66 under anthropogenic warming.

67 Reproductive strategies in plants are shaped by phylogenetic history. Fruit type, for instance,  
68 shows strong conservatism: in a dataset of 9,370 species from China, phylogeny explained  
69 nearly 80% of the variation in whether species produced fleshy or dry fruits, far exceeding the  
70 influence of growth form or climate (Wang *et al.*, 2022). Seed size is also rooted in evolutionary  
71 history. An analysis of nearly 13,000 species showed that the largest divergences in seed mass  
72 were repeatedly associated with major phylogenetic splits (Moles *et al.*, 2005). Patterns of  
73 interannual variation in reproduction also carry a phylogenetic signal (Pearse *et al.*, 2020; Qiu  
74 *et al.*, 2023). In a synthesis of 517 species, the coefficient of variation of seed production, a  
75 measure of masting intensity, exhibited significant coherence across the plant tree of life, with

76 lineages such as Fagales, Pinales, and Poales showing consistently high variability (Journé *et al.*,  
77 2023). Together, these findings indicate that reproductive strategies are constrained by ancestry.  
78 Consequently, woody plants are expected to be evolutionary constrained in the magnitude of seed  
79 production and growth potential, resulting in weaker or stronger allocation trade-offs depending  
80 on lineage (Pearse *et al.*, 2020; Journé *et al.*, 2023; Dorken *et al.*, 2025).

81 Functional traits capture physiological and morphological strategies that affect fitness indi-  
82 rectly through their influence on growth, survival, and reproduction (Violle *et al.*, 2007; Adier  
83 *et al.*, 2014). Because traits are relatively easily measurable and comparable between taxa, they  
84 allow generalizations about life-history strategies across species and ecosystems (Díaz *et al.*,  
85 2016; Maynard *et al.*, 2022). Analyses of tropical tree demography reveal that functional traits  
86 align with two axes of life-history variation (Rüger *et al.*, 2018; Kambach *et al.*, 2022). The  
87 fast–slow continuum, reflecting the growth–survival trade-off, and a stature–recruitment axis,  
88 which distinguishes tall, long-lived pioneer species with large seeds and low recruitment from  
89 short-lived breeders that produce numerous small seeds at the expense of growth and survival  
90 (Rüger *et al.*, 2018; Kambach *et al.*, 2022). On the other hand, nutrient-demanding species  
91 with high foliar nitrogen, high specific leaf area (SLA), and low wood density tend to produce  
92 many small seeds, whereas conservative species with dense wood, low foliar nitrogen, and  
93 low SLA produce fewer but larger seeds (Bogdziewicz *et al.*, 2023). In trees, the classic seed  
94 size–number trade-off is not strict, and species with large seeds often produce a greater total  
95 biomass of seeds (Qiu *et al.*, 2022). These results yield contrasting expectations: conservative,  
96 dense-wooded species with large seeds may experience high reproductive allocation and thus  
97 greater costs to growth (Qiu *et al.*, 2022; Bogdziewicz *et al.*, 2023). Alternatively, according  
98 to the stature–recruitment axis, tall species with large seeds may incur lower reproductive costs  
99 because they invest relatively little in recruitment while maintaining performance.

100 Spatial variation in temperature and precipitation often mediates reproductive allocation and  
101 growth, thereby potentially influencing the apparent strength of trade-offs between them (Hulshof  
102 *et al.*, 2012; Shestakova *et al.*, 2021; Hacket-Pain *et al.*, 2018). Across biomes, reproductive  
103 allocation tends to increase from cold to warm regions (Journé *et al.*, 2022; Ward *et al.*, 2025).  
104 Within species, seed production is also often higher in warmer parts of species' ranges. For  
105 example, North American tree fecundity peaks in the warm and moist southeastern United States  
106 (Sharma *et al.*, 2022). Parallel patterns emerge for growth. Trees in the tropics grow twice as  
107 fast as in temperate and boreal biomes (Locosselli *et al.*, 2020), and within species growth  
108 is generally faster at warmer sites (Perret *et al.*, 2024), except where water limitation prevails  
109 (Klesse *et al.*, 2024). Similarly, low precipitation and high evaporative demand are associated  
110 with stronger negative correlations between growth and reproduction (Hulshof *et al.*, 2012;  
111 Hacket-Pain *et al.*, 2017). Together, these findings suggest that accounting for local climate is  
112 needed to understand growth–reproduction trade-offs: warm and moist conditions may produce  
113 apparent positive associations between growth and reproduction, whereas hot and dry climates  
114 are expected to strengthen trade-offs through resource limitation.

115 Climate change affects resource allocation to reproduction and growth (Etzold *et al.*, 2020;  
116 Hacket-Pain & Bogdziewicz, 2021; Clark *et al.*, 2021). Direct effects of warming and moisture  
117 limitation often reduce both seed production and growth (Sharma *et al.*, 2022; Klesse *et al.*,  
118 2024; Perret *et al.*, 2024), yet plant responses in many systems appear to maintain reproduction  
119 at the expense of growth (Dohrenbusch *et al.*, 2002; Rowland *et al.*, 2018; Bogdziewicz *et al.*,  
120 2020; Hacket-Pain *et al.*, 2025). In European beech, warming has shifted reproduction towards  
121 more frequent seed crops, depleting stored reserves and causing a 28% decline in mean radial  
122 growth (Hacket-Pain *et al.*, 2025). That change was associated with a weakening of the nega-  
123 tive growth–reproduction correlation because, in poor seed years, trees with depleted reserves  
124 cannot mount strong growth responses (Hacket-Pain *et al.*, 2025). Experimental evidence sup-  
125 ports this mechanism, for example, *Picea abies* maintained allocation to reproduction at the  
126 expense of growth under experimental drought in Germany (Hesse *et al.*, 2021). Alternative  
127 mechanisms can also relax trade-offs when resource inputs increase through nitrogen and CO<sub>2</sub>  
128 fertilization. Rising atmospheric CO<sub>2</sub> enhances photosynthetic productivity and increases both  
129 wood production and reproductive output (e.g., FACE experiments on oaks) (Norby *et al.*, 2024;  
130 Esquivel-Muelbert *et al.*, 2025; Jablonski *et al.*, 2002). Long-term nitrogen addition often  
131 increases aboveground wood biomass and seed production (Magill *et al.*, 2004; Bogdziewicz  
132 *et al.*, 2017). Anthropogenic change can alter not only the magnitude of demographic rates but  
133 also the relationships among them, with particularly strong indirect effects expected in masting  
134 species where interactions among resource reserves, growth, and intermittent large seed crops  
135 are pronounced (Sala *et al.*, 2012; Han *et al.*, 2017; Hacket-Pain *et al.*, 2017).

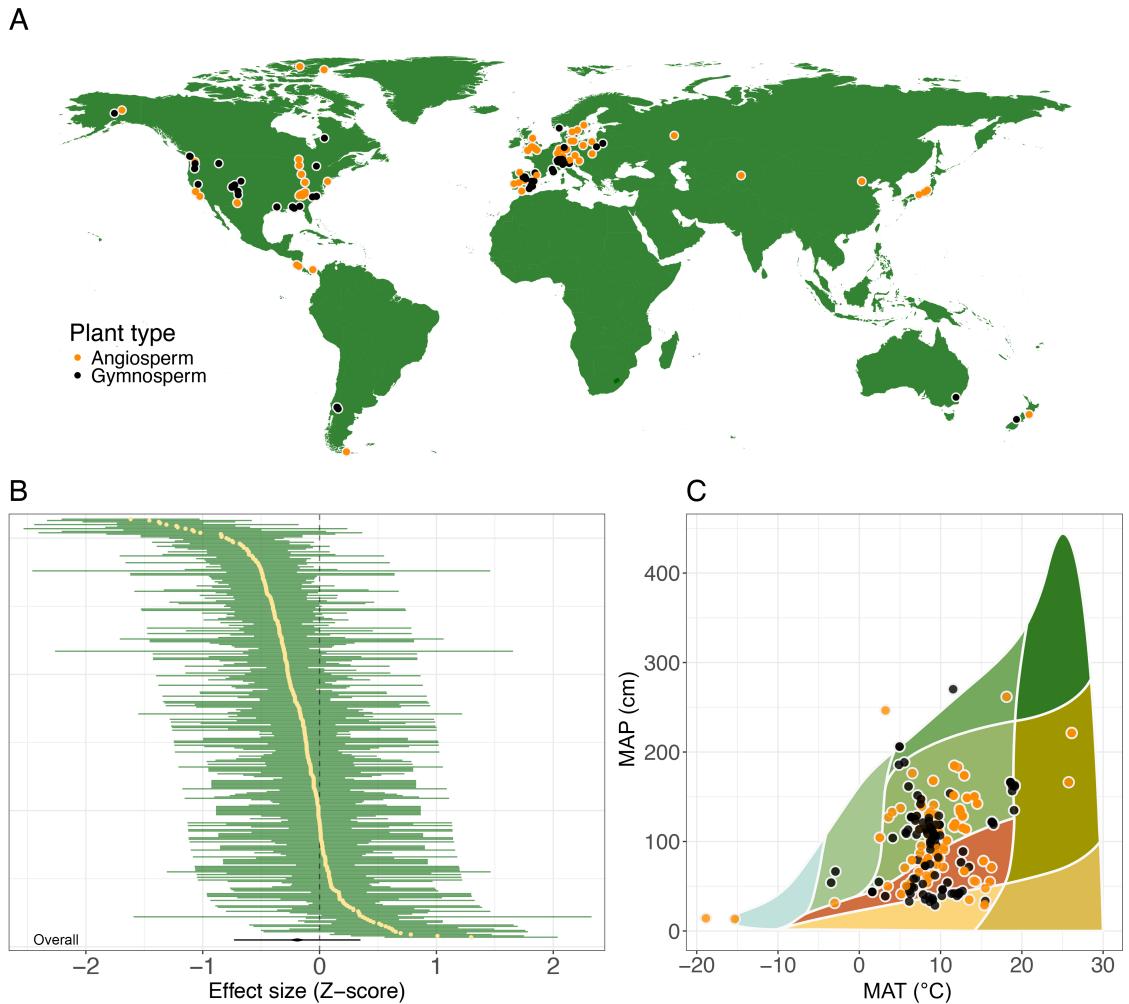
136 Interannual variation in reproductive allocation raises the question of whether growth–reproduction  
137 trade-offs operate within years or emerge across years (Knops *et al.*, 2007; Pearse *et al.*, 2016).  
138 Resource-budget models of masting predict that reproduction draws on reserves accumulated in  
139 previous years, while reserve replenishment after mast events can compete with other carbon  
140 sinks such as growth (Isagi *et al.*, 1997; Kabeya *et al.*, 2017, 2021; Kelly *et al.*, 2025). Because  
141 reproduction is temporally autocorrelated (Koenig *et al.*, 2003; Foest *et al.*, 2025), high repro-  
142 ductive output in a given year generates a null expectation of reduced reproduction—and thus  
143 relatively higher growth—both in the preceding and the following year. Under this expectation,  
144 elevated growth in the year preceding and following mast events simply reflects low reproduc-  
145 tive investment and associated growth release. Failure to observe elevated growth in the year  
146 following mast events represents a deviation from that null expectation and is consistent with  
147 delayed costs of reproduction expressed after reproductive investment. Empirical support for  
148 such lagged costs is mixed: some studies report reduced subsequent growth or storage following  
149 reproduction (Obeso, 2002; Sala *et al.*, 2012), whereas others find that trade-offs are largely  
150 confined within the same year (Hadad *et al.*, 2021; Hacket-Pain *et al.*, 2017). Explicitly ac-  
151 counting for null expectations imposed by reproductive autocorrelation is therefore important  
152 for correctly interpreting temporal growth–reproduction relationships and for linking allocation  
153 dynamics with demographic performance (Hacket-Pain *et al.*, 2018).

154 To disentangle these mechanisms, we compiled evidence from studies that jointly measured  
155 interannual seed production and aboveground, somatic growth across woody plants. We tested  
156 five hypotheses related to a broad context of factors that can influence trait trade-offs, including  
157 phylogenetic history, functional trait syndromes, spatial variation, and climate-driven effects. (1)  
158 Carry-over effects: growth-reproduction trade-offs are strongest within years, but negative effects  
159 on growth are expected in the year following high reproductive investment. (2) Phylogeny: trade-  
160 off strength is phylogenetically structured, with some lineages showing consistently stronger  
161 allocation conflicts. (3) Traits: functional traits mediate allocation patterns; species with  
162 conservative syndromes (dense wood, low SLA) exhibit higher reproductive allocation (Qiu  
163 *et al.*, 2022; Bogdziewicz *et al.*, 2023), so they should pay larger growth costs reflected in  
164 stronger negative growth-reproduction correlations. Alternatively, tall species with large seeds  
165 should display weaker trade-off correlations under the stature-recruitment axis of life history  
166 variation (Rüger *et al.*, 2018; Kambach *et al.*, 2022). (4) Climate: the trade-off depends on  
167 local climate, weakening under warm-moist conditions that promote resource accumulation  
168 and strengthening under hot-dry conditions that increase carbon and water limitation. (5)  
169 Reproductive variability and time: growth-reproduction trade-offs are stronger in species with  
170 high interannual variability in seed production (high  $CV_p$ ), reflecting large, episodic reproductive  
171 investments that impose greater demands on resources (Norton & Kelly, 1988), but weaken  
172 over time under anthropogenic change as more frequent seed years and rising resource inputs  
173 (elevated  $CO_2$ , nitrogen deposition) decouple annual reproductive effort from short-term growth  
174 responses.

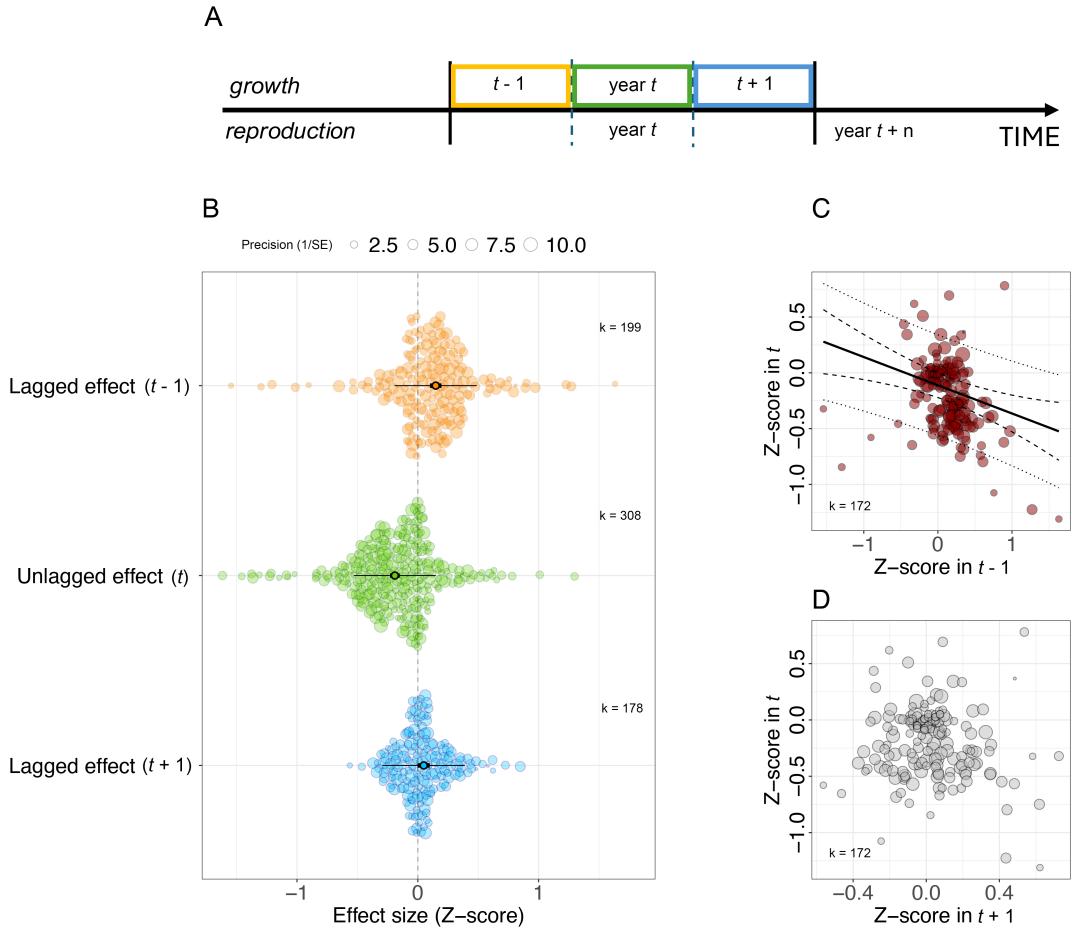
## 175 Results

176 Our meta-analysis of 78 studies, 79 species (Fig. 1), and 685 effect sizes across time lags  
177 shows that the strength and sign of growth-reproduction correlations depend on their temporal  
178 alignment (Fig. 2A,B, Supplementary Table S3). When growth and reproduction were corre-  
179 lated in the same year, the pooled effect size was negative (effect size =  $-0.194$ , 95% CI =  
180  $[-0.239, -0.148]$ ,  $N = 308$ ), consistent with the within-year trade-offs between growth and  
181 reproduction. Hereafter, we refer to negative growth-reproduction correlations as trade-offs,  
182 following their common interpretation as signatures of allocation constraints, while acknowl-  
183 edging that the underlying evidence is correlational (see Discussion).

184 Higher growth in year  $t - 1$  was associated with greater reproduction in year  $t$  (effect  
185 size =  $0.143$ , 95% CI =  $[0.090, 0.197]$ ,  $N = 199$ ), consistent with the expectation that years  
186 preceding high reproductive output tend to be characterized by low reproductive investment  
187 and, thus, relatively higher growth (Fig. 2B). This association was strongest in populations with  
188 stronger within-year trade-offs (Fig. 2C, Supplementary Table S4), consistent with reproductive  
189 autocorrelation generating strong growth contrasts between years of reproductive suppression  
190 and preceding low-investment years; particularly in populations with strong growth-reproduction



**Figure 1: Data distribution and effect size.** (A) The location of population-level observations of growth-reproduction correlations. (B) Population-level effect sizes (Z-scores, see Methods) for the relationship between reproduction and growth in the same year ( $t$ ) with associated 95% confidence intervals (for lagged effects and their framework see Fig. 2A,B). The yellow points indicate the mean of each effect size. The estimated effect size of the null model with the confidence and prediction intervals is shown below ( $k = 308$  observations and  $n_{\text{species}} = 79$ ). (C) Whittaker biome plot with observations distributed along mean annual temperature [ $^{\circ}\text{C}$ ] and mean annual precipitation [cm]. Orange points indicate species belonging to angiosperms, while black points indicate species belonging to gymnosperms. Background colors refer to the Whittaker Biome classification ranging from tundra (pale blue) through temperate forests in the middle to tropical seasonal and tropical rain forest (olive and dark green).

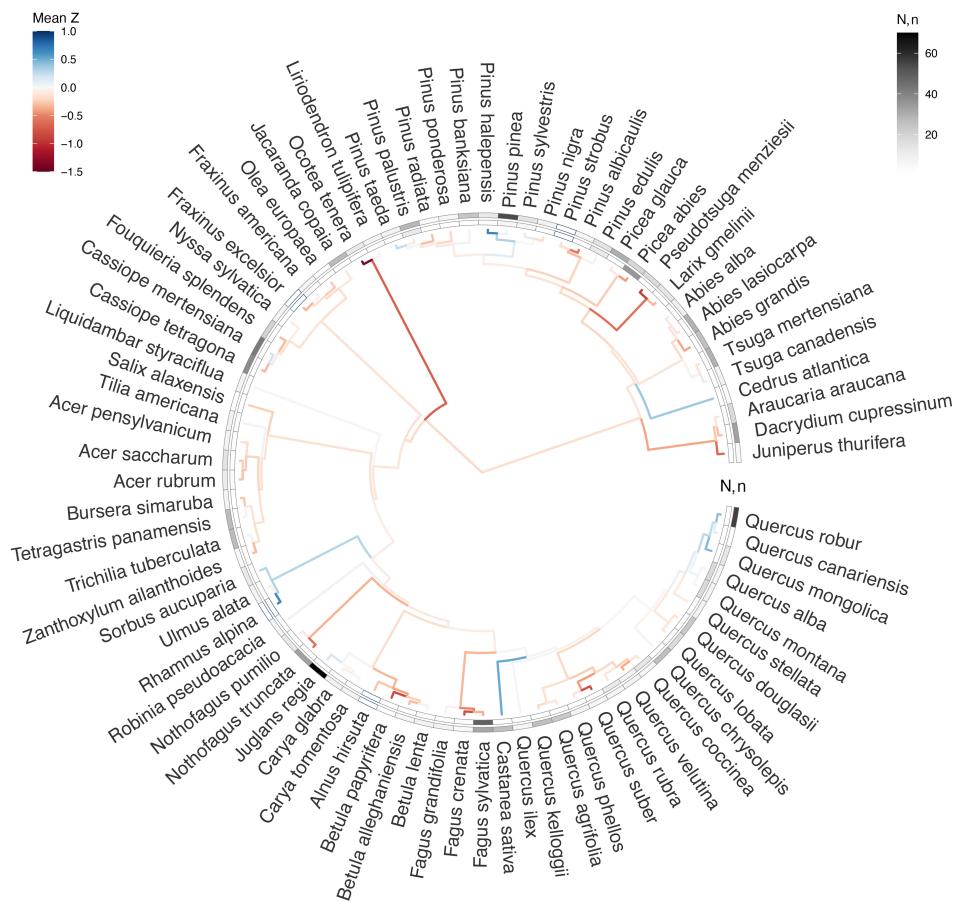


**Figure 2: Growth is reduced in years of high seed production.** (A) The temporal framework to study growth-reproduction correlations across multiple years, accounting for the within-year and lagged effect of growth ( $t-1$ ,  $t$ , and  $t+1$  in relation to reproduction in year  $t$ ). (B) Relationships between reproduction and growth in the preceding year ( $t-1$ ), same year ( $t$ ), and the subsequent year ( $t+1$ ). Black points show estimated effect sizes with 95% confidence intervals (CI; thicker bars) and 95% prediction intervals (PI; thinner bars). The relationship between the growth-reproduction correlation (Z-score) in the same year ( $t$ ) and C) reproduction and growth in the preceding year ( $t-1$ ), and D) reproduction and growth in the following year ( $t+1$ ). Line at C) shows the relationship as estimated with phylogenetically informed multilevel meta-analytic model, associated 95% confidence intervals (dashed line), and 95% prediction intervals (dotted line). Points represent population-level effect sizes ( $k$ ), with point size proportional to their precision (inverse SE). Model summary provided in Supplementary Table S3.

191 trade-offs. In contrast, reproduction in year  $t$  was not associated with growth in year  $t+1$   
 192 (effect size = 0.041, 95% CI = [-0.016, 0.099],  $N = 178$ ) (Fig. 2B). The absence of a  
 193 positive association contrasts with the expectation of elevated growth in years adjacent to high  
 194 reproductive output, indicating an asymmetry around reproductive events. Such a pattern is  
 195 consistent with delayed costs of reproduction expressed through constrained post-reproductive  
 196 growth.

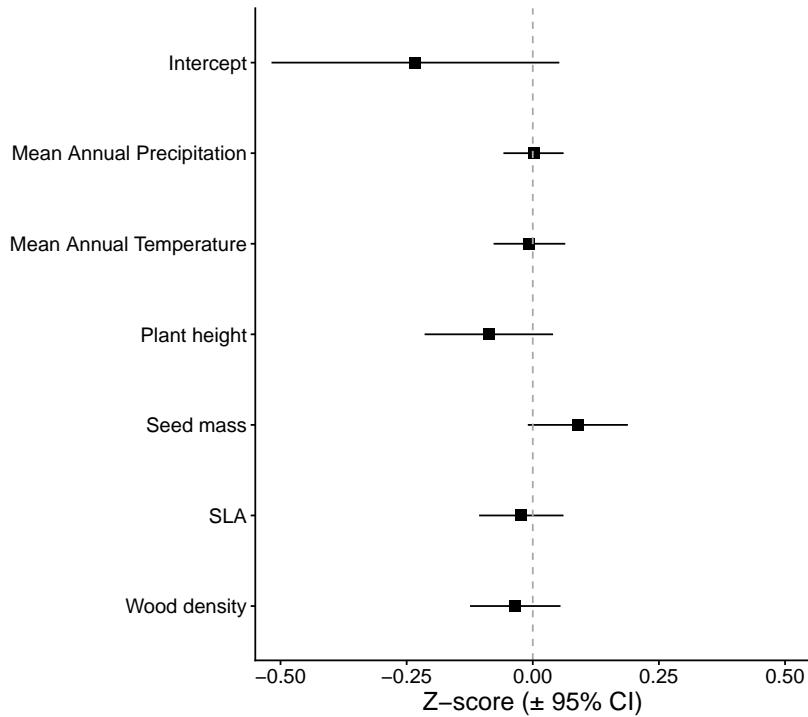
197 The growth-reproduction trade-off (within-year/unlagged effect) based on the effect sizes  
 198 averaged for each of the 79 species was not significantly related to phylogeny (Pagel's  $\lambda = 0.001$ ,  
 199  $p = 0.99$ , Fig. 3). Thus, closely related species do not have more similar effect sizes than  
 200 expected by chance. This lack of signal reflects the heterogeneous distribution of effect size  
 201 across the phylogeny: positive and negative effects occur intermixed within most major clades,

202 with no large clusters of uniformly positive or negative values. Even within genera for which  
 203 a larger number of species were sampled, such as *Quercus*, *Pinus*, and *Fagus*, the direction  
 204 and magnitude of effects vary widely among species. Some adjacent taxa differ sharply in  
 205 effect size, indicating contrasting growth–reproduction correlations despite close evolutionary  
 206 relatedness. Within oaks, for example, *Quercus robur* had a positive effect size on average,  
 207 while *Q. suber* or *Q. velutina* negative. Similarly, the negative effect sizes in *Fagus sylvatica*  
 208 and *F. crenata*, contrasted with near-zero correlation in *F. grandifolia*. Closely related *Pinus*  
 209 *nigra* and *P. sylvestris* were characterized by negative and positive effect sizes, respectively. The  
 210 weak phylogenetic structure arises from repeated, independent expression of both negative and  
 211 positive correlations between growth and reproduction across the plant tree of life.



**Figure 3: Growth–reproduction correlation does not exhibit phylogenetic coherence.** Red indicate negative, and blue indicate positive correlations between reproduction and growth in the same year ( $t$ ). Phylogenetic signal, estimated from species-level mean effect sizes, is negligible (Pagel’s  $\lambda = 0.001$ ,  $p = 0.99$ ,  $n_{\text{species}} = 79$ ). The gray-scale circle indicates sampling intensity, with shade proportional to the number of populations sampled (inner ring,  $N$ ) and mean number of years per species (outer ring,  $n$ ) for each species.

212 While we expected the negative correlation between growth and reproduction to be weaker  
 213 in warm-moist sites, the interaction between mean annual temperature (MAT) and mean annual  
 214 precipitation (MAP) was not significant (estimate = 0.003,  $p = 0.9$ ). The trade-off did not  
 215 vary systematically across sites differing in MAT ( $p = 0.62$ ) and MAP ( $p = 0.63$ ) (Fig. 4,



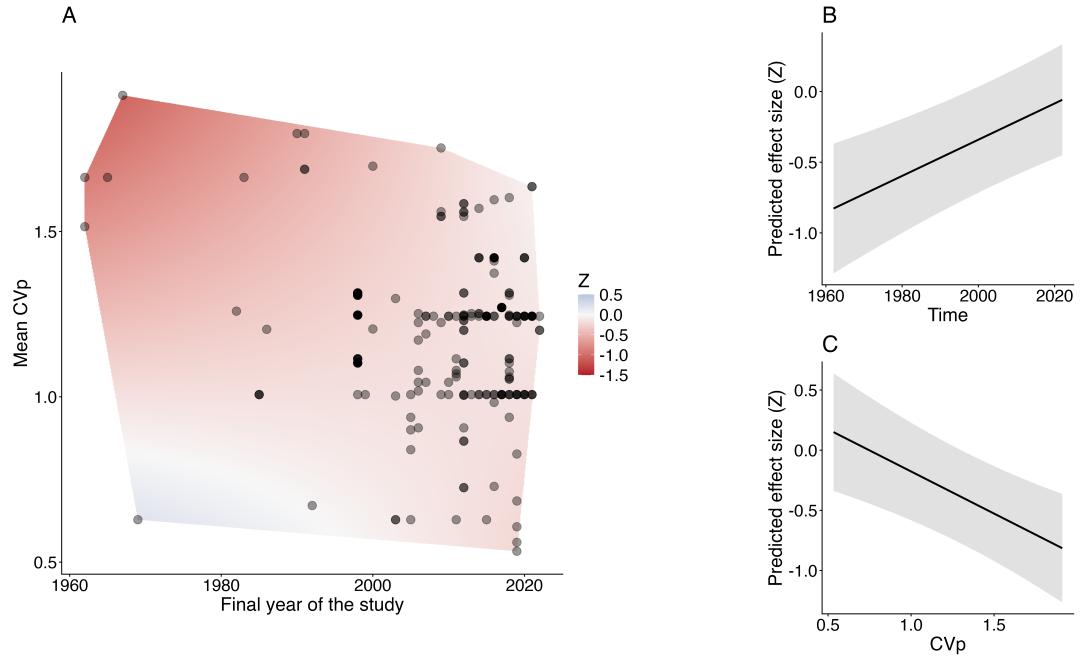
**Figure 4: Growth-reproduction correlation shows no associations with climate or functional traits.** Effect sizes and their 95% confidence intervals are derived from a phylogenetically informed multilevel comparative model ( $n_{\text{observations}} = 307$ ,  $n_{\text{species}} = 79$ ), with sampling variance incorporated at the observation level. All predictors were centered and scaled to facilitate direct comparison of effect sizes. Model summary provided in Supplementary Table S6.

216 Supplementary Table S6).

217 Wood density ( $p = 0.37$ ) and specific leaf area (SLA,  $p = 0.85$ ), two key traits reflecting  
 218 the fast-slow continuum, were not significantly related to the trade-off between growth and  
 219 reproduction (Supplementary Table S6). Likewise, the maximum height, which is reflecting  
 220 the stature-recruitment trade-off, was not associated with the trade-off between growth and  
 221 reproduction ( $p = 0.17$ ). Seed size, the key trait related to reproduction, exhibited only a weak  
 222 positive tendency to moderate growth-reproduction trade-off ( $p = 0.06$ ), such that larger seed  
 223 sizes were associated with weaker trade-offs.

224 The correlations between growth and reproduction became weaker over time, especially for  
 225 species with high interannual variation in reproduction (year  $\times$   $\text{CV}_p$  interaction:  $\beta = 0.021$ ,  
 226  $p < 0.001$ ,  $n = 261$ ) (Fig. 5, Supplementary Table S7). For example, in a species with an  
 227 intermediate  $\text{CV}_p = 1$ , the estimated correlation between growth and reproduction was -0.16 in  
 228 both 1980 and 2020. In contrast, in a species characterized by higher interannual variation in  
 229 reproduction ( $\text{CV}_p = 1.6$ ), the estimated correlation declined more than 8-fold over the 40 years,  
 230 from -0.58 in 1980 to -0.07 in 2020 (Fig. 5B).

231 At the same time, in earlier decades of the dataset, species with high  $\text{CV}_p$  in seed produc-  
 232 tion exhibited stronger negative growth-reproduction correlations than species with low  $\text{CV}_p$   
 233 (Fig. 5C). For example, in species with high reproductive variability ( $\text{CV}_p = 1.6$ ), the esti-  
 234 mated correlation was -0.8, whereas in species with relatively regular reproduction ( $\text{CV}_p = 0.6$ )



**Figure 5: Negative growth-reproduction correlations weaken over time in strongly masturing species.** The surface plot at A) shows the estimated (unlagged) growth-reproduction correlation across combinations of population-level interannual variation in seed production ( $CV_p$ ) and the final monitoring year of the study from which each effect size was extracted. Black dots denote data points that define the convex hull of the prediction space ( $n_{\text{populations}} = 261$ ,  $n_{\text{species}} = 48$ ). B) Temporal weakening of negative correlation between growth and reproduction, as estimated for  $CV_p = 1.6$ . C) The negative correlation between growth and reproduction is stronger in species characterized by higher  $CV_p$ , as estimated for year 1980. All predictions are derived from phylogenetically informed multilevel meta-analytic model, summarized in Supplementary Table S7.

235 the corresponding estimate was 0.27. Over time, however, growth-reproduction correlations  
 236 converged across levels of reproductive variability.

## 237 Discussion

238 Our meta analysis indicate a trade-off between growth and reproduction in woody plants: growth  
 239 and seed production are negatively correlated within the same year. In addition, the temporal  
 240 structure of these correlations reveals asymmetric carry-over effects across years, consistent  
 241 with delayed costs of reproduction expressed as constrained growth following high reproductive  
 242 investment. The trade-off exhibits no detectable phylogenetic signal. Although the absence of  
 243 signal contrasts with our predictions, the result aligns with demographic syntheses indicating that  
 244 allocation strategies are only weakly constrained by lineage (Salguero-Gómez *et al.*, 2016; Rüger  
 245 *et al.*, 2018). Furthermore, neither spatial variation in climate nor functional traits systematically  
 246 explained variation in effect sizes, suggesting that species resolve allocation conflicts in ways  
 247 that are not tightly linked to biome, climatic regime, or ecological strategy. Notably, the trade-off  
 248 was strongest in species with high interannual variability in reproduction but diminished over  
 249 recent decades in these species, suggesting that changing resource dynamics under warming are  
 250 reshaping the balance between growth and reproduction (Hacket-Pain *et al.*, 2025; Macias &

251 Redmond, 2025). Our results indicate that while growth–reproduction trade-offs are widespread,  
252 their expression is flexible and responsive to contemporary environmental change (summarized  
253 in Supplementary Table 1).

254 Within-year trade-offs dominated growth–reproduction dynamics, but lagged correlations  
255 revealed a pronounced asymmetry around reproductive events. Growth and seed production  
256 were negatively correlated in the same year. Growth in year  $t - 1$  was positively associated with  
257 reproduction in year  $t$ , a pattern that scaled with the strength of within-year growth suppression.  
258 This scaling indicates that the pre-reproductive growth signal primarily reflects large contrasts  
259 between years when growth is strongly suppressed by reproductive investment and preceding  
260 years when reproduction is minimal (Koenig *et al.*, 2003; Foest *et al.*, 2025). In contrast,  
261 reproduction in year  $t$  was not associated with growth in year  $t + 1$ . Given the temporal  
262 autocorrelation of reproduction, the absence of elevated growth in the post-reproductive year  
263 contrasts with the expectation of low reproductive investment and indicates constrained growth  
264 following reproduction. These results support resource-budget models in which reproduction  
265 draws on resources accumulated in previous years and replenishment after mast events competes  
266 with growth, such that delayed costs are expressed as asymmetric temporal patterns around  
267 reproductive events rather than as a consistent lagged reduction in growth (Satake & Iwasa,  
268 2000; Han *et al.*, 2017; Kelly *et al.*, 2025). This interpretation, by adjusting the null expectation  
269 to account for the autocorrelation in reproduction, contrasts with earlier interpretations of positive  
270 growth anomalies at  $t - 1$  indicating resource accumulation, or a lack of growth response at  $t + 1$   
271 as indicating an absence of delayed costs of reproduction (Hacket-Pain *et al.*, 2017; Drobyshev  
272 *et al.*, 2010; Nussbaumer *et al.*, 2021).

273 The lack of phylogenetic signal in the growth–reproduction correlation contrasts with the  
274 evolutionary conservatism reported for reproductive traits such as seed size, fruit type, and  
275 interannual variability in seed production (Moles *et al.*, 2005; Wang *et al.*, 2022; Journé *et al.*,  
276 2023). While lineages such as Fagales and Pinales show coherent patterns of masting intensity  
277 (Journé *et al.*, 2023), or conserved seed size (Moles *et al.*, 2005), our results indicate that the  
278 strength and even the sign of growth–reproduction correlations vary markedly among closely  
279 related species. Positive and negative correlations occur side by side within genera including  
280 *Quercus*, *Fagus*, *Pinus*, and *Tsuga*, implying repeated divergence in allocation strategies within  
281 clades. This suggests that, although key reproductive traits are constrained by ancestry, the  
282 balance between growth and reproduction remains evolutionarily labile.

283 We also found no evidence that spatial variation in climate modulates the trade-off. Neither  
284 mean annual temperature, mean annual precipitation, nor their interaction predicted variation in  
285 the growth–reproduction correlation. This result contrasts with the expectation that warm–moist  
286 sites should weaken trade-offs by easing resource limitation, while hot–dry climates should  
287 strengthen them. Several mechanisms may explain this mismatch. First, coarse annual climate  
288 metrics may not capture the within-season water stress and extreme events that drive resource  
289 limitation or sink activity (Espelta *et al.*, 2008; Körner *et al.*, 2023; Cabon, 2025). Second,

**Table 1: Summary of results across hypotheses.** Hypotheses are stated in shortened form; results summarize the direction and support in the meta-analysis.

Hypothesis	Result summary
Within-year allocation trade-off	Growth and reproduction are negatively correlated within the same year, indicating a widespread immediate cost of reproduction.
Lagged costs of reproduction	Growth is higher in the year before reproduction but not elevated after reproduction, consistent with delayed post-reproductive growth constraints.
Phylogenetic constraint	Growth–reproduction correlations show no phylogenetic signal; closely related species often differ strongly in trade-off strength and sign.
Trait mediation	Functional traits (wood density, SLA, height) do not explain variation in trade-off strength; seed mass shows at most a weak tendency toward weaker trade-offs.
Climatic modulation	Trade-off strength is not systematically related to mean annual temperature or precipitation across sites.
Reproductive variability ( $CV_p$ )	Species with high interannual reproductive variability exhibit stronger negative growth–reproduction correlations.
Temporal change	Negative growth–reproduction correlations weaken over recent decades, particularly in species with high reproductive variability.

290 growth and reproduction may respond in parallel to local climate, so that their correlation is  
 291 conserved even when absolute rates change (Locosselli *et al.*, 2020; Ward *et al.*, 2025). Third,  
 292 local edaphic variation and stand structure can buffer or amplify climatic effects (Barringer *et al.*,  
 293 2013), reducing the signal of broad-scale climate drivers in meta-analytic models.

294 Similarly, functional traits associated with the fast–slow continuum and the stature–recruitment  
 295 axis did not show associations with the growth–reproduction correlation. Wood density, spe-  
 296 cific leaf area, and maximum height were all poor predictors of trade-off strength, and seed  
 297 size showed only a weak, marginally significant tendency toward more positive correlations.  
 298 This suggests that simple trait syndromes, while powerful for predicting demographic rates and  
 299 fecundity (Rüger *et al.*, 2018; Qiu *et al.*, 2022; Bogdziewicz *et al.*, 2023), may not map directly  
 300 to how plants partition resources between growth and reproduction from year to year, especially  
 301 under variable environmental conditions (Augusto *et al.*, 2025). In fact, a recent global analysis  
 302 failed to detect the expected links between acquisitive trait syndromes and high growth rates in  
 303 natural forests, as environmental constraints override trait-based growth potential across species  
 304 (Augusto *et al.*, 2025). Trade-offs may instead be governed by unmeasured axes of variation,  
 305 such as non-structural carbohydrate storage dynamics (Trugman & Anderegg, 2025) or pheno-  
 306 logical strategies (Journé *et al.*, 2021; Etzold *et al.*, 2021; Wolkovich *et al.*, 2025), which are  
 307 rarely quantified across many species but may determine how plants buffer or transmit resource  
 308 limitations into growth and seed output.

309 The temporal trend in our dataset reveals that in strongly masting species, the negative  
310 growth–reproduction correlation has weakened over recent decades. First, the strong relationship  
311 between the interannual variation in reproduction ( $CV_p$ ) and growth-reproduction trade-off  
312 supports an early, but so far untested, prediction that high variation in reproductive allocation in  
313 strongly masting species should render the detection of costs of reproduction more likely in such  
314 species (Norton & Kelly, 1988). The weakening of the trade-off over time aligns with the long-  
315 term evidence from European beech and Japanese oak (*Quercus crispula*), in which warming  
316 increased the frequency of weather cues that trigger reproduction, leading to more frequent seed  
317 crops, and in the case of beech, reserve depletion, and an erosion of the growth–reproduction  
318 trade-off (Foest *et al.*, 2024; Hacket-Pain *et al.*, 2025; Shibata *et al.*, 2020). Our results indicate  
319 that this weakening of the growth–reproduction trade-off is not unique to beech, implying  
320 a more general disruption of how trees balance competing demands on limited resources.  
321 Such disruption matters because growth–reproduction correlations reflect underlying resource  
322 allocation processes. Warming can push these processes out of balance, with consequences that  
323 extend from growth and reproduction to survival, defense, and overall resilience (Lauder *et al.*,  
324 2019; Macias & Redmond, 2025). A weakening trade-off therefore does not necessarily signal  
325 relaxed resource limitation. Rather, it may indicate that trees are increasingly unable to convert  
326 the opportunity for compensatory growth in years of reduced reproductive investment because  
327 reserves are depleted or physiological stress constrains recovery (Hacket-Pain *et al.*, 2025). In  
328 strongly masting species, this creates conditions where reproduction remains frequent while  
329 growth declines — a combination shown to reduce carbon storage, diminish future reproductive  
330 potential, and increase mortality (Qiu *et al.*, 2021; Hacket-Pain *et al.*, 2025; Bordin *et al.*, 2025).  
331 These temporal shifts suggest that climate change is altering the resource dynamics that underpin  
332 masting, weakening a fundamental component of demographic regulation in many forest trees.

333 Our findings come with several limitations. While the database include a large geographic  
334 range, it is largely composed of temperate and boreal tree species, reflecting the current scarcity  
335 of long-term seed production data for tropical systems (Hacket-Pain *et al.*, 2022). While alloca-  
336 tion strategies and environmental constraints may differ under such conditions, community-level  
337 studies suggest reduced growth during general flowering events in tropical Dipterocarp forests  
338 (Nakagawa *et al.*, 2012). Similarly, a larger species sample with broader trait and climate cov-  
339 erage could reveal patterns that were not detected here, despite the reasonable trait coverage  
340 of our dataset (Supplementary Figure S5). Measures of growth and reproduction also vary  
341 across studies—ranging from basal area increment and ring width to fruit counts and seed  
342 traps—introducing methodological heterogeneity that our models only partly capture (Sup-  
343 plementary Table S5). Potentially important sources of variation, such as ontogenetic stage  
344 (Thomas, 2011), stand age (Genet *et al.*, 2009; Ward *et al.*, 2025), or competitive environment  
345 (Clark *et al.*, 2014; Augusto *et al.*, 2025), were often unreported and could not be incorpo-  
346 rated despite their possible influence on allocation patterns. Our analysis also largely focuses on  
347 secondary growth of the main stem, whereas primary growth may be more directly affected by re-

348 productive investment, especially given the partial resource autonomy of branches (Hoch, 2005;  
349 Han *et al.*, 2011). Negative growth–reproduction relationships may also arise not from direct  
350 competition for shared resources but from independent sink limitation by external drivers—for  
351 example, weather conditions that favor reproduction while suppressing growth (Knops *et al.*,  
352 2007; Mund *et al.*, 2020). Growth can also be actively down-regulated in years with high  
353 reproductive investment to maintain storage (Dietze *et al.*, 2014; Trugman & Anderegg, 2025).  
354 While such mechanisms could contribute to the patterns we detect, the meta-analysis was not  
355 designed to disentangle them. A further limitation is that most studies rely on among-year cor-  
356 relations averaged across individuals, even though trade-offs evolve and operate at the level of  
357 individual plants (Cople *et al.*, 2021). Many estimates of trade-offs therefore reflect aggregated  
358 data on growth and reproduction from different trees; if resource allocation depends strongly  
359 on individual-level resource acquisition, such aggregation could obscure or inflate underlying  
360 relationships. Finally, the 'decline effect' in meta-analyses, caused by an increasing tendency  
361 to publish null results (Nakagawa *et al.*, 2022), may overestimate temporal effects on trade-offs  
362 caused by a changing climate. Together, these limitations highlight the need for long-term  
363 datasets that jointly track individuals, multiple tissues, demographic stages, and environmental  
364 covariates to fully resolve how growth–reproduction relationships operate across species and  
365 biomes.

366 Taken together, our results show that growth–reproduction trade-offs in woody plants are  
367 common and weakly constrained by lineage, climate, or functional traits, yet are changing over  
368 time in strongly masting species. The consistent same-year cost of reproduction, combined  
369 with asymmetric lagged effects in which high reproductive investment is followed by reduced  
370 growth, supports a resource-budget view in which allocation conflicts are expressed within  
371 years and lagged costs emerge following high reproductive investment (Satake & Iwasa, 2000;  
372 Koenig *et al.*, 2003). The absence of strong phylogenetic, climatic, or trait controls indicates  
373 that trade-off strength is not a fixed property of lineages or life-history syndromes, but emerges  
374 from local conditions and species-specific resource dynamics (Hacket-Pain *et al.*, 2017). The  
375 temporal weakening of the trade-off in high- $CV_p$  species links these dynamics to climate change,  
376 suggesting that more frequent cueing and altered reserve trajectories can decouple growth  
377 responses from annual reproductive effort (Hacket-Pain *et al.*, 2025). Our meta-analysis calls for  
378 further research on resource-based individual-level quality of phenotypic correlations between  
379 growth and reproduction in woody plants in relation to the changing climate.

## 380 Materials and Methods

381 **Meta-analysis** To identify relevant studies, we conducted keyword searches in Web of Science  
382 and Scopus following the PRISMA guidelines (Preferred Reporting Items for Systematic Reviews  
383 and Meta-Analyses); (O'Dea *et al.*, 2021). We evaluated three different combinations of terms  
384 related to growth and reproduction (Supplementary Table S1) inspired by two systematic reviews  
385 on the cost of reproduction in plants (Obeso, 2002; Thomas, 2011) and the referenced studies  
386 incorporated in our meta-analysis. We applied the final search string on 30 March 2024 including:

387 (reproduction OR seed OR seeds OR fruit\* OR cone\* OR acorn\*) AND ("tree  
388 growth" OR ring\*) AND (tree\* OR shrub\*)

389 Given the large number of studies reporting growth (>10,000, Supplementary Table S1) in  
390 non-woody plants, we restricted the search to woody species by using "tree growth" instead of  
391 "growth". We also limited the search to trees and shrubs by including these terms directly in the  
392 string.

393 The search yielded 3,449 and 3,111 documents from Web of Science and Scopus, re-  
394 spectively. To streamline the process of literature screening we used Rayyan QCRI (<https://rayyan.ai/>). After removing duplicates (2,015), we screened abstracts of 4,545 articles,  
395 and classified 110 papers for full-text evaluation. That classification was based on a previ-  
396 ously constructed decision tree inspired by studies from two systematic reviews (Supplementary  
397 Figure S1). We also included data from before-mentioned systematic reviews (Obeso, 2002;  
398 Thomas, 2011) and studies cited in recent publications, resulting in a total of 146 publications  
399 eligible for data extraction (Supplementary Figure S2). Studies on orchard trees and cultivated  
400 species were excluded due to the possible bias caused by horticultural selection (Supplementary  
401 Figure S1).

403 Out of the 146 articles reporting both annual reproduction and annual growth in woody  
404 plants, we positively classified 78 studies based on assessment of the full text. Sixty eight  
405 studies that did not contain reliable data on either growth or reproduction, or did not report  
406 appropriate statistical tests between them were excluded (68 articles in total). In case of studies  
407 lacking formal statistical analyses or raw data suitable for our analysis, we contacted 14 authors  
408 asking for the missing data. As five research groups shared their data on six species, we were  
409 able to include additional data representing over 60 tree populations. We extracted all lagged  
410 effects of reproduction on growth when reported.

411 We extracted data combining somatic growth and reproduction over multiple years, with a  
412 few exceptions for studies based on one year of measurements. Most studies used secondary  
413 growth and seed production, but in the final analysis, we also included 13 studies on primary  
414 growth and physiological measurements, and 6 studies that measured flower production instead  
415 of seed production (Supplementary Figure S3). Excluding these studies (based on less frequent  
416 categories of growth or reproduction measurement) did not change the qualitative outcome of the

417 analyses despite statistically significant differences in effect sizes among growth measurement  
418 methods (Supplementary Figure S4, Supplementary Table S5).

419 In total, we compiled a dataset of 685 observations (species  $\times$  population  $\times$  time-lag effect)  
420 from 78 studies containing data on reproduction and growth of trees and other woody plants.  
421 We also extracted available effect sizes that were reported in the studies, but were not included  
422 in the final analysis, e.g. time lags exceeding one year, correlations based on flower numbers or  
423 female individuals only (70 extra effect sizes). Given that seed production in some species (e.g.  
424 *Pinus pinea* (Garcia-Barreda *et al.*, 2021)) extends over multiple years we classified correlations  
425 between time lags in respect to the species' natural history. For example, for *Pinus pinea*, *Pinus*  
426 *palustris* and *Cedrus atlantica* we chose the year of seed filling as the ecologically relevant year  
427 to be correlated with growth.

428 To standardize data for the meta-analysis, we first transformed all reported effect sizes to  
429 Pearson's correlation coefficient ( $r$ ) using test-specific formulas (Supplementary Table S2). We  
430 then converted  $r$  to Fisher's Z-scores ( $Z_r$ ) using

$$Z_r = \frac{1}{2} \ln\left(\frac{1+r}{1-r}\right).$$

431 The sampling variance of each effect size was calculated as

$$V_{Z_r} = \frac{1}{n-3}.$$

432 For most observations (97%), the sample size ( $n$ ) was equivalent to the number of years for  
433 which reproduction and growth were correlated. In case of growth-reproduction comparisons  
434 that used different statistical tests from correlations, we used the number of years or the number  
435 of individuals as the respective sample size. If the sample size was less than four, the observation  
436 was excluded from the analysis.

437 **Climate data** We extracted mean annual temperature (MAT [ $^{\circ}$ C]) and mean annual precip-  
438 itation (MAP [cm]) for each unique study location, based on monthly values of temperature  
439 and precipitation (1970 to 2000) from the corresponding 2.5 minute grid cell in the WorldClim  
440 2 dataset (Fick & Hijmans, 2017). The extracted values of MAT and MAP were plotted in a  
441 Whittaker biome diagram using the *plotbiomes* package (Stefan & Levin, 2025).

442 **Functional traits and masting metrics** We compiled functional trait data from publicly  
443 available datasets (Kattge *et al.*, 2020; Chave *et al.*, 2009). We filtered the list of species  
444 according to the World Checklist of Vascular Plants (WCVP) (Brown *et al.*, 2023). Then  
445 we compiled data on species-specific functional traits related to the fast-slow continuum and  
446 defined by the global trait analysis (Díaz *et al.*, 2016; Reich, 2014). We selected specific leaf  
447 area (SLA), wood density, seed mass, and plant height as potential predictors of the growth-

448 reproduction trade-off. We obtained complete set of functional traits for 61 species. For each  
449 trait x species combination that was absent from the databases, we averaged the trait value at  
450 the genus level. Such a trait imputation allowed us to cover all 79 species with the relevant trait  
451 information. We log-transformed all trait values to approximate a normal distribution, and to  
452 facilitate comparisons, we scaled and centered these values prior to the analysis.

453 To obtain the coefficient of variation of seed production ( $CV_p$ ), we filtered MASTREE+,  
454 a global database of annual records of population-level reproductive effort (Hacket-Pain *et al.*,  
455 2022; Foest *et al.*, 2024). We estimated the species-level  $CV_p$  by taking at least 5-year time series  
456 from at least 5 distinct populations for each species. We were able to extract the mean  $CV_p$  of  
457 seed production for 48 species. If information on population-specific seed production variability  
458 ( $CV_p$ ) was reported in the study (11 species), we gave it priority over the MASTREE-derived  
459 value.

460 **Analysis** All data analysis was performed in R v. 4.5.1 (R Core Team, 2024). We used  
461 the multilevel comparative meta-analysis approach that involves phylogenetic, interspecific, and  
462 intraspecific variance partitioning (Pottier *et al.*, 2024). We built our models using the *rma.mv*  
463 function in the *metafor* package v. 4.8-0 (Viechtbauer, 2010). First, based on our species  
464 list, we built the phylogenetic tree using the Open Tree of Life with the *rotl* package v. 3.1.0  
465 (Michonneau *et al.*, 2016). We calculated the distance between branches using Grafen's method  
466 and then built the correlation matrix of phylogenetic relatedness under Brownian motion with  
467 the *ape* package v. 5.8-1 (Paradis & Schliep, 2019). All fitted models had the same random  
468 effect structure that included effects of individual study, individual effect size, and phylogenetic  
469 relatedness. In addition to this model structure, we built two parallel models for sensitivity  
470 analyses, one using the phylogenetic relatedness adapted from Zanne *et al.* (Zanne *et al.*, 2014)  
471 and one simplified model replacing the phylogenetic relatedness correlation matrix with just the  
472 random effect of species. We used *orchaRd* package v. 2.0 to plot the results from the *rma*  
473 models (Nakagawa *et al.*, 2023).

474 To test whether growth-reproduction trade-offs occur in woody plants worldwide, we ran a  
475 multilevel comparative model with time lags included as the model predictors ( $n = 685$ ). The  
476 models tested whether effect sizes (Fisher's Z) differed from zero for correlations measured in  
477 year  $t$  (a), one-year lags  $t - 1$  (b), and one-year lags  $t + 1$  (c) (Fig. 2A). We also quantified the  
478 relationship between growth and reproduction correlation within-year with the lagged correla-  
479 tions. To do so, we fitted an *rma* model using the within-year Z-scores as response, and the  
480 lagged correlations ( $t - 1$  and  $t + 1$ ) as predictors.

481 In further analyses, we restricted the dataset to within-year correlations ( $t$ ; a;  $n = 308$ ). We  
482 then tested for a phylogenetic signal in the trade-off expression. To this end, we averaged the  
483 Z-score at the species level and then calculated Pagel's  $\lambda$  (Paradis & Schliep, 2019).

484 To assess how climate and functional traits modulate growth-reproduction trade-offs, we  
485 fitted a multilevel comparative model. Predictors included mean annual temperature (MAT) and

486 mean annual precipitation (MAP) as key climatic variables, and specific leaf area (SLA), seed  
487 mass, plant height, and wood density as relevant functional traits.

488 Finally, we fitted a model to test how trade-offs in species characterized by variable  $CV_p$   
489 of seed production have changed over recent decades. To do so, we extracted the final year of  
490 the study for each observation and fitted a model that included the final year of the study, the  
491 average species-specific  $CV_p$  of seed production, and their interaction as predictors.

492 For each model, we estimated the total amount of heterogeneity ( $I^2$ ) to quantify variation  
493 that is not explained by sampling error. The information on heterogeneity was given in model  
494 summaries together with the estimated variance components of random effects terms (See  
495 Supplementary Material).

496 We assessed potential publication bias using two methods recommended by Nakagawa *et al.*  
497 (2022). Specifically, we plotted a funnel plot with Z-scores and performed multilevel meta-  
498 regressions with standard error and sampling variance as fixed predictors of effect size. Visual  
499 inspection of the funnel plot suggests that there is no publication bias (Supplementary Figure S6).  
500 Likewise, the meta-regressions revealed that there is no bias related to studies with small sample  
501 sizes as both sampling variance and standard error were non-significant predictors of the effect  
502 size (Fisher's Z).

503 **Acknowledgements**

504 We would like to acknowledge researchers who kindly shared their data: Sabine Braun, Simon  
505 Tresch, Sven Hopf, Angelika Kölbl, Burkhard Beudert, Bruno Fady, Maxime Cailleret, Manuel  
506 Esteban Lucas Borja, and Walt Koenig. This study was funded by Polish National Science Cen-  
507 tre grant 2019/35/D/NZ8/00050 and the European Union (ERC, ForestFuture, 101039066; EC,  
508 FECUND, 101244227). Views and opinions expressed are however those of the authors only  
509 and do not necessarily reflect those of the European Union or the European Research Council.  
510 Neither the European Union nor the granting authority can be held responsible for them. JJF  
511 was also supported by the Foundation for Polish Science (FNP).

512

513 **Author Contributions Statement**

514 AHP, MB and MKB conceived the study idea and developed research questions. MKB, MB,  
515 UE, MB, VJ, KK, JS and AHP performed the abstract screening. MKB extracted data from the  
516 literature and curated data. MKB led the analysis with the input from SMD, MB, VJ, JJF and  
517 JS. MKB wrote the first draft of the manuscript with substantial help from MB and AHP. All  
518 authors contributed to the interpretation of the analysis, revised the draft, and gave their final  
519 approval for publication.

520

521 **Declaration of interests**

522 Authors have no competing interests to declare.

523

524 **Data availability statement**

525 The data and code supporting the results are available at [https://github.com/maciejkbarczyk/meta\\_trade-offs](https://github.com/maciejkbarczyk/meta_trade-offs).

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803 **Supplementary Information**

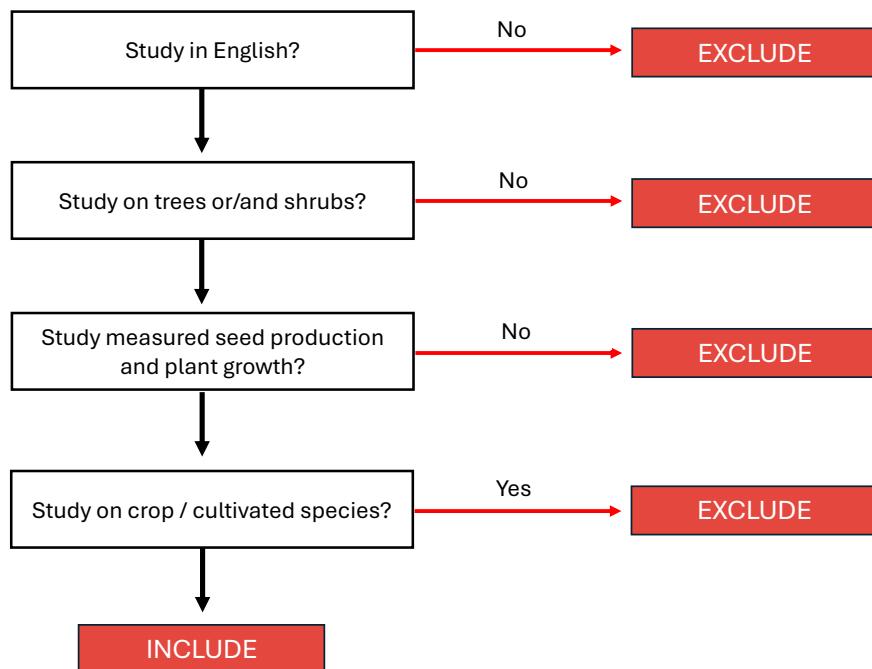
804 **Title:** Growth–reproduction trade-offs are common but changing in woody plants: a meta-  
805 analysis

806 **Authors:** Maciej K. Barczyk, Michał Bogdziewicz, Szymon M. Drobniak, Maria Bogdańska,  
807 Urszula Eichert, Jessie J. Foest, Valentin Journé, Katarzyna Kondrat, Jakub Szymkowiak, An-  
808 drew Hacket-Pain

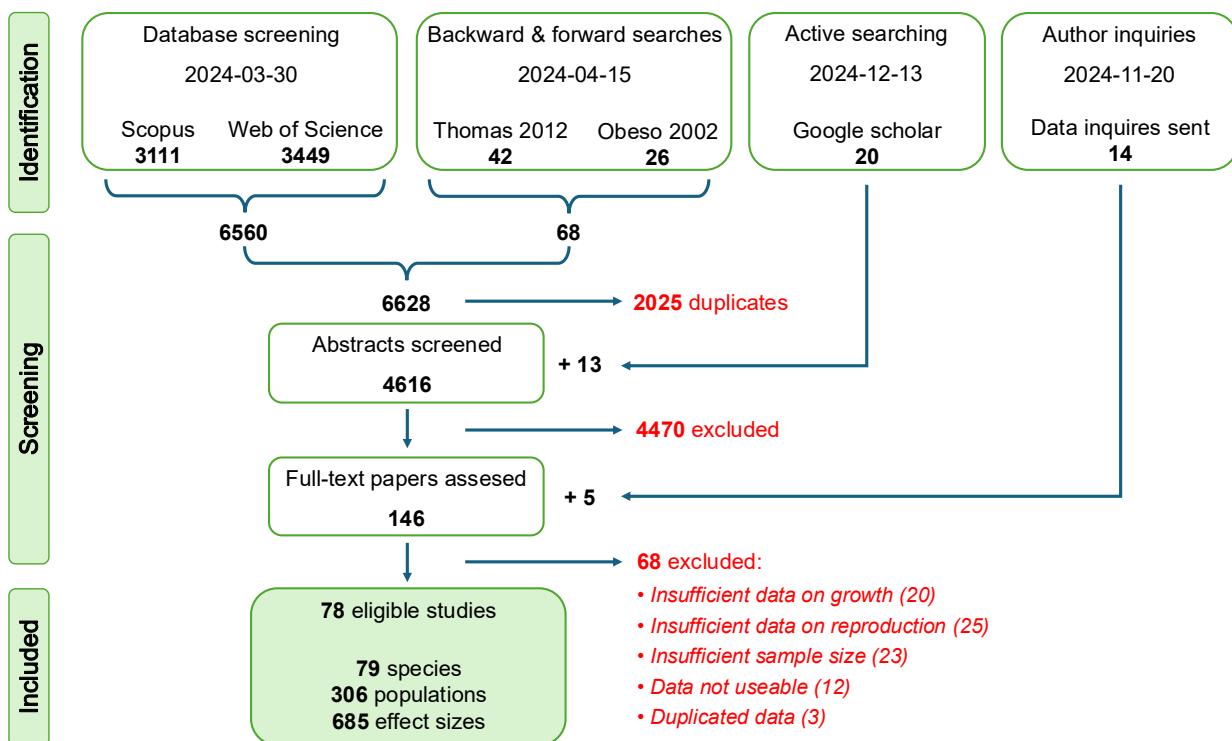
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**Table S1: Search summary.** Each row contains set of keywords used for searching in the databases and the numbers of research items detected in the respective database. The trial search was performed at 22.03.2024. The final string was rerun at 30.03.2024 and supplied the meta-analysis (see Methods).

String	Scopus	Web of Science
(reproduction OR seed* OR cone* OR acorn*) AND (growth OR ring*) AND tree	23,244	22,733
(reproduction OR seed* OR cone* OR acorn* OR fruit*) AND (growth OR ring*) AND tree	33,863	32,367
(reproduction OR seed OR seeds OR fruit* OR cone* OR acorn* ) AND ("tree growth" OR ring* ) AND ( tree* OR shrub*)	3,102	3,253



**Figure S1: Decision tree.** Decision tree with the inclusion criteria used in meta-analysis.



**Figure S2: PRISMA flow chart.** PRISMA graph summarizes the methods and data used in meta-analysis, including the search description, number of screened and excluded studies, the reasons for exclusion, and the final details on eligible studies.

**Table S2: Effect size formulas.** Statistical tests and equations used to calculate standardized Pearson's r correlation coefficient. Intermediate effect size (t-value) was calculated with the use of respective effect size (ES) and standard error (SE) prior to conversions to Pearson's r coefficients following  $t = ES/SE$ . The n refers to the sample size.

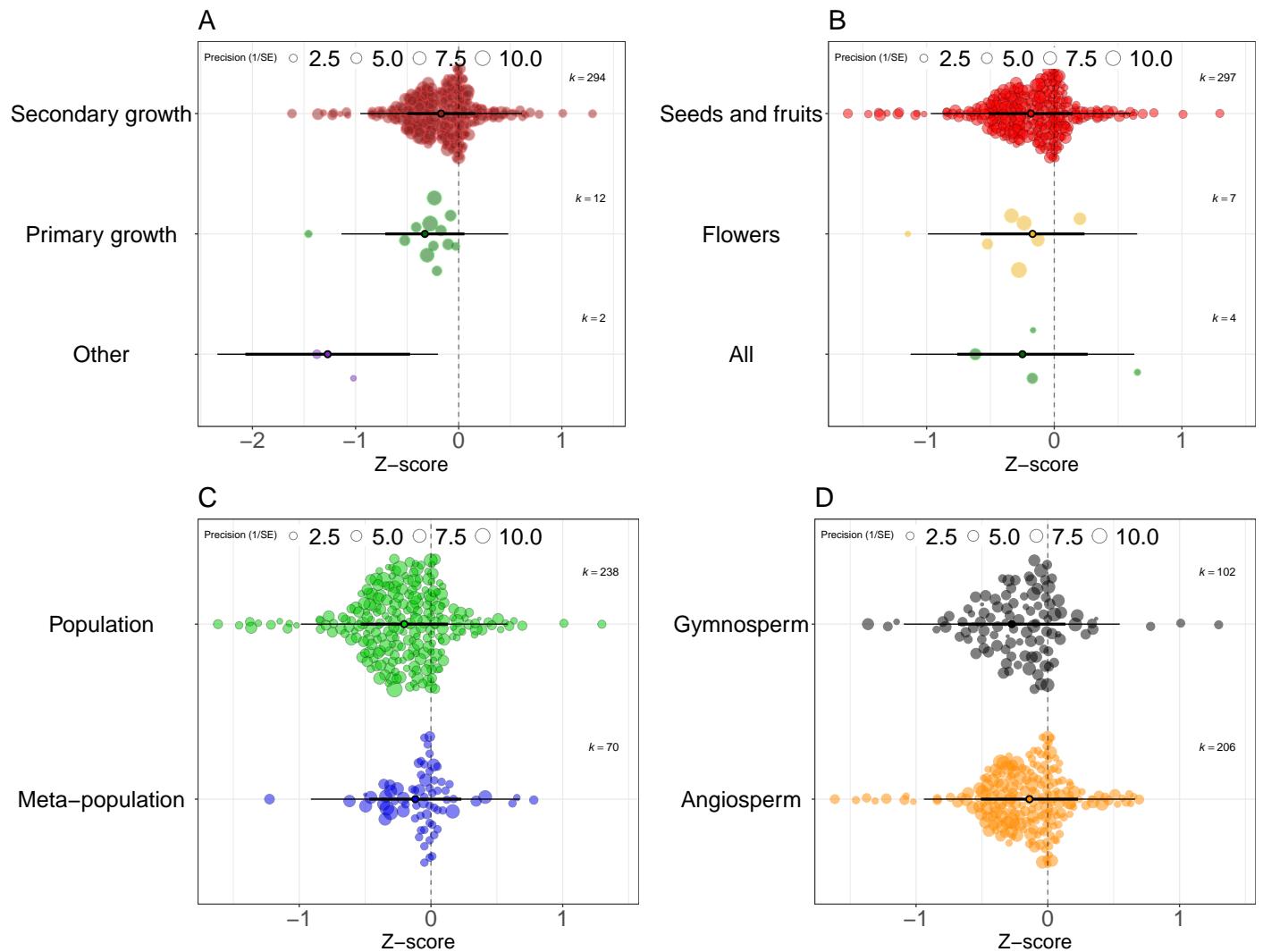
Statistical test	Formula
<b>t-test</b>	$r_P = \sqrt{t^2/(t^2 + n - 1)}$
<b>Spearman correlations</b>	$r_P = 2 \sin(\pi r_S/6)$
<b>Linear models</b>	$r_P = \sqrt{(t^2/(t^2 + n - 1))}$
<b>Multiple regressions</b>	$r_P = \sqrt{t^2/(t^2 + n - 2)}$

**Table S3: Results of the rma model testing the relationships between reproduction and growth in the preceding year ( $t-1$ ), same year ( $t$ ), and the subsequent year ( $t+1$ ).** The effect sizes are based on phylogenetically informed multilevel meta-analytic model without an intercept accounting for random effects of phylogeny, effect size ID, and study. The results are based on 685 observations and 79 species [2](#). Asterisks indicate significant effects. AIC = 443.63;  $I^2 = 47.11$ . The estimated variance components of random effects equal <0.001 for phylogeny, 0.01 for study, and 0.03 for effect size ID.

Predictor	estimate	se	z-value	p-value	ci.lb	ci.ub
Lagged effect ( $t+1$ )	0.039	0.03	1.304	0.192	-0.02	0.097
Unlagged effect ( $t$ )	-0.199	0.024	-8.388	<.0001	-0.245	-0.152 ***
Lagged effect ( $t-1$ )	0.143	0.028	5.120	<.0001	0.088	0.198 ***

**Table S4: Summary of the rma model testing the relationship between the growth-reproduction correlation (Z-score) in the same year ( $t$ ), and reproduction and growth in the preceding year ( $t - 1$ ), and reproduction and growth in the following year ( $t + 1$ ) (Fig. 2).** The effect sizes are based on phylogenetically informed multilevel meta-analytic model accounting for random effects of phylogeny, effect size ID, and study. The results are based on 172 observations and 51 species for which information on population-level correlation was available. Asterisks indicate significant effects. AIC = 26.02;  $I^2 = 48.82$ . The estimated variance components of random effects equal <0.001 for phylogeny, 0.05 for study, and <0.001 for effect size ID.

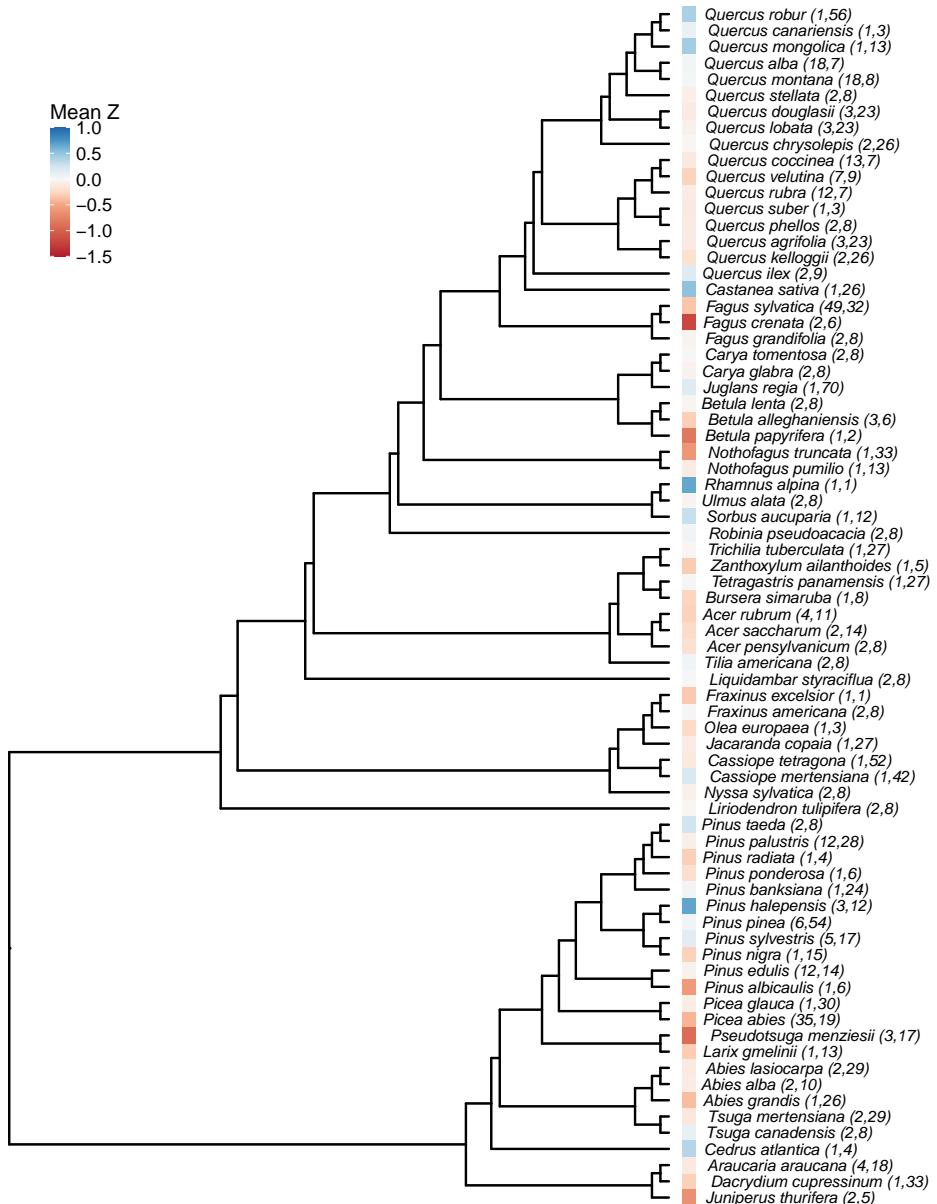
Predictor	estimate	se	z-value	p-value	ci.lb	ci.ub
Intercept	-0.107	0.056	-1.924	0.054	-0.2161	0.002
Lagged effect ( $t - 1$ ) Z	-0.253	0.079	-3.219	0.001	-0.407	-0.099
Lagged effect ( $t + 1$ ) Z	-0.166	0.098	-1.702	0.089	-0.357	0.025



**Figure S3: Category-dependent differences in population-level, within-year growth-reproduction correlations.** A) Comparison of the effect sizes (Z-score) between different growth measurements. Secondary growth measurements derive mostly from tree ring data and basal area increments. Primary growth refers to measurements of buds, twigs, organs or plant height increments. Other methods ("others") refer to two studies using photosynthetic capacity [O evolution/leaf area] and root collar non-structural carbohydrate (NSC) levels as growth proxy. B) Comparison of the effect sizes (Z-score) between different reproduction measurements. C) Comparison of the effect sizes (Z-score) between populations and meta-populations. D) Comparison of the effect sizes (Z-score) between gymnosperms and angiosperms. Model summary provided in Supplementary Table S5.

**Table S5: Summary of the rma model testing the growth-reproduction relationships in relation to the different data categories.** The effect sizes are based on the phylogenetically informed multilevel meta-analytic model accounting for random effects of phylogeny, effect size ID, and study. Primary and secondary growth were compared with other growth measurements (three-level factor); flowers, seeds and fruits were compared with all reproductive organs (three-level factor); populations were contrasted with meta-populations (two-leveled factor) and gymnosperms were compared with angiosperms (two-leveled factor) in a joint rma model, as visualized in the Supplementary Fig. S3. The results are based on 308 observations derived from 78 studies on 79 species. Asterisks indicate significant effects. AIC = 194.11;  $I^2 = 69.08$ . The estimated variance components of random effects equal 0.08 for phylogeny, 0.05 for study, and <0.001 for effect size ID.

Predictor	estimate	se	z-value	p-value	ci.lb	ci.ub	
Intercept	-1.226	0.46	-2.667	0.008	-2.127	-0.325	**
Primary growth	0.942	0.402	2.343	0.019	0.154	1.73	*
Secondary growth	1.1	0.39	2.82	0.005	0.335	1.864	**
Population	-0.084	0.084	-1.005	0.315	-0.248	0.08	
Flowers	0.079	0.243	0.326	0.744	-0.397	0.556	
Seeds and fruits	0.066	0.204	0.323	0.747	-0.335	0.467	
Gymnosperms	-0.134	0.212	-0.633	0.527	-0.549	0.281	



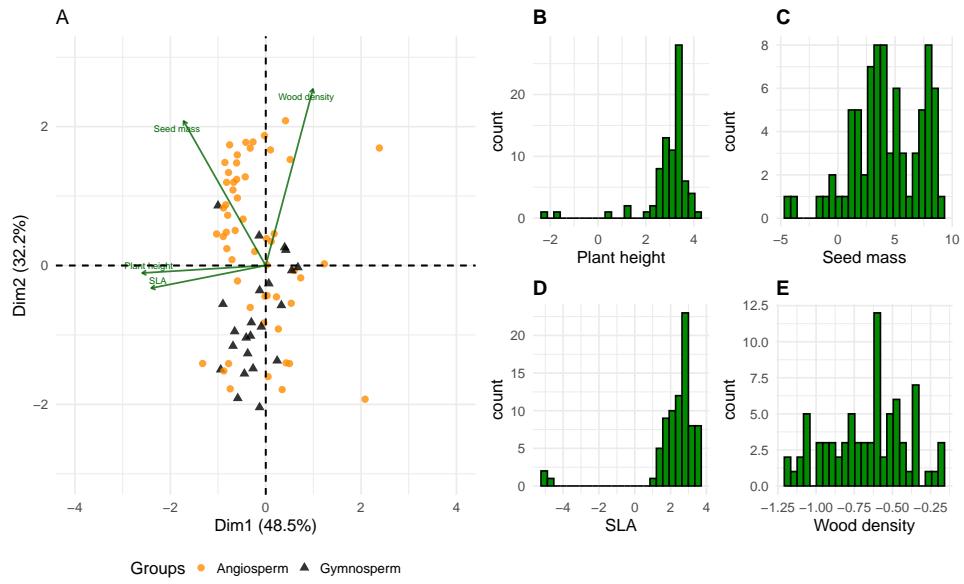
**Figure S4: Correlation between secondary growth and reproduction does not exhibit phylogenetic coherence.** Warmer colours (reds) indicate negative, and cooler colours (blues) indicate positive correlations between reproduction and growth in the same year ( $t$ ). In brackets given are: the number of populations in which trade-offs were studied and the mean number of years for each species. Phylogenetic signal, estimated from species-level mean effect sizes, is negligible (Pagel's  $\lambda < 0.001$ ,  $p = 0.99$ ,  $n = 74$  species). The data is restricted to observations based only on studies measuring secondary growth (see Supplementary Fig. S3).

**Table S6: Summary of the rma model testing the effects of climate and functional traits on growth-reproduction relationships (Fig. 4).** The effect sizes are based on phylogenetically informed multilevel meta-analytic model accounting for random effects of phylogeny, effect size ID, and study. MAT - Mean Annual Temperature, MAP - Mean Annual Precipitation, SLA - Specific Leaf Area. The results are based on 308 observations and 79 species. AIC = 202.09;  $I^2 = 65.76$ . The estimated variance components of random effects equal 0.06 for phylogeny, 0.05 for study, and <0.001 for effect size id.

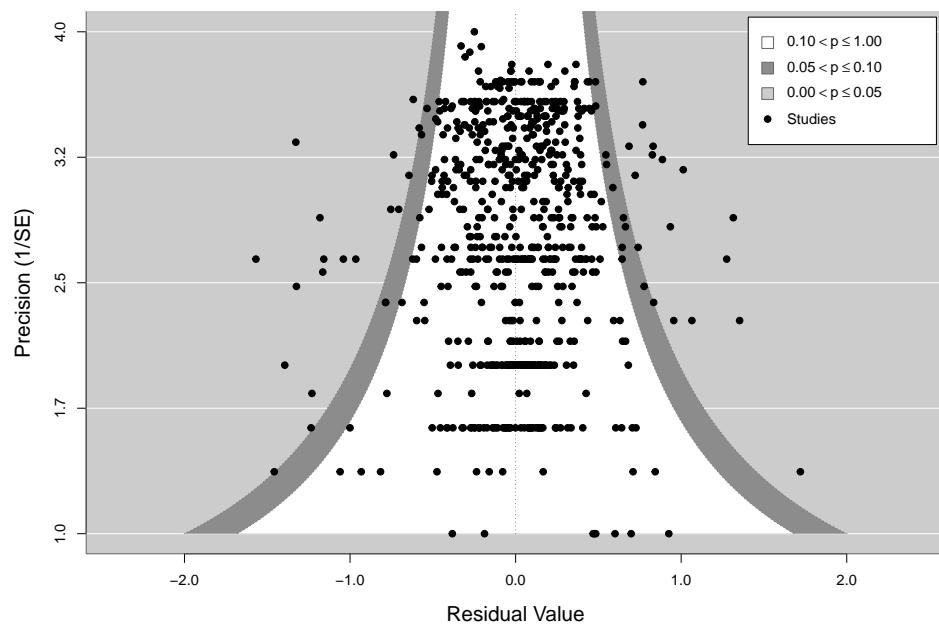
Predictor	estimate	se	z-value	p-value	ci.lb	ci.ub
Intercept	-0.233	0.146	-1.599	0.11	-0.518	0.053
MAT	-0.007	0.036	-0.183	0.855	-0.078	0.064
MAP	0.002	0.03	0.048	0.962	-0.058	0.061
SLA	-0.023	0.043	-0.534	0.594	-0.106	0.061
Wood density	-0.035	0.046	-0.758	0.446	-0.124	0.055
Plant height	-0.087	0.065	-1.342	0.18	-0.214	0.04
Seed mass	0.089	0.051	1.765	0.077	-0.01	0.189

**Table S7: Summary of the rma model testing the effect of interaction between mean  $CV_p$  of seed production and the final year of the study.** The effect sizes are based on phylogenetically informed multilevel meta-analytic model accounting for random effects of phylogeny, effect size ID, and study. Asterisks indicate significant effects. The results are based on 261 observations and 48 species for which  $CV_p$  values were available (Fig 5). Asterisks indicate significant effects.  $AIC = 142.61$ ;  $I^2 = 64.02$ . The estimated variance components of random effects equal 0.09 for phylogeny, 0.02 for study, and  $<0.001$  for effect size id.

Predictor	estimate	se	z-value	p-value	ci.lb	ci.ub	
Intercept	42.64	15.9	2.682	0.007	11.477	73.803	**
Mean $CV_p$	-42.778	12.148	-3.522	0.0004	-66.587	-18.969	***
Final year of the study	-0.021	0.008	-2.687	0.007	-0.037	-0.006	**
Mean $CV_p$ : Final year	0.021	0.006	3.512	0.0004	0.009	0.033	***



**Figure S5: Coverage of species-level functional traits investigated in the meta-analysis.** A) Principal Component Analysis. Arrow length indicate relationships between four functional traits tested in the meta-analysis (i.e. SLA, plant height, wood density, seed mass) (Fig. 4) averaged at the species level ( $n = 79$ ). Each point represents single species (yellow - angiosperms, black triangles - gymnosperms). B) Histogram of log-transformed plant height. C) Histogram of log-transformed seed mass. D) Histogram of log-transformed SLA (specific leaf area). E) Histogram of log-transformed wood density (including genus-level average values).



**Figure S6: Funnel plot.** Residual values from the rma model are plotted in relation to their precision (inverse standard error). Each point refers to population-level growth-reproduction correlation (based on all time lags,  $n = 685$ ). Shades highlight different levels of statistical significance described in the legend.