1 One to rule them all? Assessing the performance of Forest

2 Europe's biodiversity indicators against multitaxonomic data

3

4 Impact statement

- 5 Biodiversity indicators used to assess the state of Europe's forests perform unequally; a
- 6 combination of indicators is more informative
- 7

8 Keywords

- 9 Sustainable forest management; taxonomic indicators; multi-taxa diversity; forest structure
- 10

11 Authors

12 Yoan Paillet, Univ. Grenoble Alpes, INRAE, Lessem, 2 rue de la Papeterie, BP76, 38402 13 Saint Martin d'Hères, cedex, France, yoan.paillet@inrae.fr, 0000-0001-7232-7844 14 15 Livia Zapponi, National Research Council, Institute of BioEconomy, Via Francesco Biasi 16 75, 38098 S. Michele all'Adige (TN) Italy, livia.zapponi@cnr.it, 0000-0003-4033-9001 17 18 **Peter Schall**, University of Göttingen, Silviculture and Forest Ecology of the Temperate 19 Zones Büsgenweg 1, 37077 Göttingen, Germany, peter.schall@forst.uni-goettingen.de, 20 0000-0003-4808-818X 21 22 Jean-Matthieu Monnet, Univ. Grenoble Alpes, INRAE, Lessem, 2 rue de la Papeterie, 23 BP76, 38402 Saint Martin d'Hères, cedex, France, jean-matthieu.monnet@inrae.fr, 0000-24 0002-9948-9891 25 26 Christian Ammer, University of Göttingen, Silviculture and Forest Ecology of the Temperate 27 Zones Büsgenweg 1, 37077 Göttingen, Germany, christian.ammer@forst.uni-goettingen.de, 28 0000-0002-4235-0135 29 30 Lorenzo Balducci, Sapienza University of Rome, Piazzale Aldo Moro, 5, 00185 Rome, Italy, 31 lorenzo.balducci@uniroma1.it, 0000-0001-9111-5679 32 33 Steffen Boch, WSL Swiss Federal Research Institute, Zürcherstrasse 111, 8903 34 Birmensdorf, Switzerland, steffen.boch@wsl.ch, 0000-0003-2814-5343 35

36 Gediminas Brazaitis, Vytautas Magnus University Agriculture Academy Studentu g. 11, 37 53361 Akademija, Lithuania, gediminas.brazaitis@vdu.lt, 0000-0003-0234-9292 38 39 Alessandro Campanaro, Council for Agricultural Research and Economics, Research 40 Centre for Plant Protection and Certification Via Lanciola 12/a, Florence, Italy, 41 alessandro.campanaro@crea.gov.it, 0000-0003-4043-6362 42 43 Francesco Chianucci, CREA - Research Centre for Forestry and Wood Viale Santa Margherita 80, Arezzo, Italy, fchianucci@gmail.com, 0000-0002-5688-2060 44 45 46 Inken Doerfler, Carl von Ossietzky Universität Oldenburg, Department Plant sociology and 47 nature Conservation Ammerländer Heerstraße 114-118 26129 Oldenbur, Germany, 48 inkendoerfler@gmail.com, 0000-0001-9624-1922 49 Markus Fischer, University of Bern, Institute of Plant Sciences Altenbergrain 21, 3013 Bern, 50 51 Switzerland, markus.fischer@ips.unibe.ch, 0000-0002-5589-5900 52 53 Marion Gosselin, INRAE, UR EFNO Domaine des Barres, 45290 Nogent-sur-Vernisson, 54 France, marion.gosselin@inrae.fr, 0000-0003-0161-5368 55 56 Martin M. Gossner, WSL Swiss Federal Research Institute, Zürcherstrasse 111, 8903 57 Birmensdorf, Switzerland, martin.gossner@wsl.ch, 0000-0003-1516-6364 58 59 Jacob Heilmann-Clausen, Center for Macroecolgy, Evolution and Climate, Globe Institute, 60 University of Copenhagen, Universitetsparken 15, 2100 København, Denmark, jheilmann-61 clausen@sund.ku.dk, 0000-0003-4713-6004 62 63 Jeňýk Hofmeister, Czech University of Life Sciences Prague, Department of Forest 64 Ecology, Faculty of Forestry and Wood Sciences Kamýcká 129, 165 00 Praha - Suchdol, Czech Republic, jenyk.hofmeister@email.cz, 0000-0002-3915-5056 65 66 Jan Hošek, Ecological Services Tichá 784/4, 268 01 Hořovice, Czech Republic, 67 68 hosek@ekologickesluzby.cz 69 70 Sebastian Kepfer-Rojas, University of Copenhagen, Department of Geosciences and 71 Natural Resource Management Rolighedsvej 23, 1958 Frederiksberg, Denmark, 72 skro@ign.ku.dk, 0000-0002-1681-2877 73 74 **Peter** Odor, Hun-Ren Centre for Ecological Research, Institute of Ecology and Botany, 75 Alkotmány u. 2-4., H-2163 Vácrátót; University of Sopron, Institute of Environmental 76 Protection and Nature Conservation, Bajcsy-Zsilinszky u. 4., H-9400 Sopron, Hungary, 77 odor.peter@ecolres.hu, 0000-0003-1729-8897 78 79 Flóra Tinya, HUN-REN Centre for Ecoogical Research, Institute of Ecology and Botany 80 Alkotmány u. 2-4., H-2163 Vácrátót, Hungary, tinya.flora@ecolres.hu, 0000-0002-4271-9676 81

82 Giovanni Trentanovi, Research Institute on Terrestrial Ecosystems - National Research 83 Council Via Madonna del Piano n. 10 50019 Sesto Fiorentino (Florence), Italy, 84 giovanni.trentanovi@cnr.it, 0000-0001-8236-4163 85 Giorgio Vacchiano, University of Milan, Department of Agricultural and Environmental 86 87 Science via Celoria 2, 20133 Milan, Italy, gvacchiano@gmail.com, 0000-0001-8100-0659 88 89 Kris Vandekerkhove, Research Institute for Nature and Forest (INBO), Gaverstraat 4, 90 9500 Geraardsbergen, Belgium, kris.vandekerkhove@inbo.be, 0000-0003-1954-692X 91 92 Wolfgang W. Weisser, Technical University of Munich, Terrestrial Ecology Research 93 Group, Department of Life Science Systems, School of Life Sciences Hans-Carl-von-94 Carlowitz-Platz 2, 85354 Freising, Germany, wolfgang.weisser@tum.de, 0000-0002-2757-95 8959 96 97 Sabina Burrascano, Sapienza University of Rome, Piazzale Aldo Moro, 5, 00185 Rome, 98 Italy, sabina.burrascano@uniroma1.it, 0000-0002-6537-3313 99 100 101 Acknowledgements 102 This work was funded by the EU Framework Programme Horizon 2020 through the COST 103 Association (www.cost.eu): COST Action CA18207: BOTTOMS-UP - Biodiversity Of

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107 Abstract

Most broad-scale forest biodiversity indicators are based on data from national forest inventories and are used to assess the state of biodiversity through several regional initiatives and reporting. Although valuable, these indicators are essentially indirect and evaluate habitat quantity and quality rather than biodiversity *per se*. Besides, most of these indicators are applicable at regional or national scales, while their use at a more local level is difficult. Therefore, their link to biodiversity may be weak, which decreases their usefulness for decision-making.

115 For several decades, Forest Europe indicators assessed the state of European forests, in 116 particular its biodiversity. However, no extensive study has been conducted to date to 117 assess the performance of these indicators against multitaxonomic data. We hypothesized 118 that – as implied by the reporting process – no single biodiversity indicator from Forest 119 Europe can represent overall forest biodiversity, but that several – eventually combined – 120 indicators would reflect habitat quality for at least some taxa in a comprehensive way. We 121 tested the set of indicators proposed by Forest Europe against the species richness of six 122 taxonomic and functional groups (tracheophytes, epixylic and epiphytic bryophytes, birds, 123 saproxylic beetles, saproxylic non-lichenized fungi and epixylic and epiphytic lichenized 124 fungi) across several hundreds of plots over Europe. We showed that, while some indicators 125 perform relatively well across groups (e.g. deadwood volume), no single indicator 126 represented all biodiversity at once, and that a combination of several indicators performed 127 better. Surprisingly, some indicators showed weak links with the biodiversity of the six 128 taxonomic and functional groups.

Forest Europe indicators were chosen for their availability and ease of understanding for most people. However, our analyses showed that there are still gaps in the monitoring framework, and that surveying certain taxa along with stand structure is necessary to support policymaking and tackle forest biodiversity loss at the large scale.

133 Keywords

134 Sustainable forest management; taxonomic indicators; multi-taxa diversity; forest structure

135 Introduction

136 Monitoring biodiversity in the face of global change is a challenge in many ecosystems 137 across the world (IPBES 2019; Lindenmayer & Likens 2010). Despite large-scale initiatives 138 such as GEOBON (Group on Earth Observation Biodiversity Observation Network 2008) and 139 collaborative networks (Burrascano et al. 2023), no standard biodiversity monitoring scheme 140 has been agreed in Europe or elsewhere. Long-term biodiversity monitoring hence currently 141 relies on a limited number of initiatives that are used to assess the impact of climate change 142 (e.g. Jiguet et al. 2012), habitat loss (Betts et al. 2017) or pollution (Rigal et al. 2023). 143 Initiatives combining monitoring of biodiversity, types of pressure and ecosystem-level 144 variables remain quite rare, or may concern only single pressure vs. single taxon 145 assessment (e.g. Proença et al. 2017; Weber et al. 2004). However, different taxonomic and 146 functional groups likely respond differently to a given pressure, which challenges 147 prioritization of policy initiatives and tools (Pereira & David Cooper 2006).

148 Forest ecosystems are no exception to the rule, despite the large use of National Forest 149 Inventory data to indirectly assess the state and evolution of biodiversity (Chirici et al. 2012; 150 Heym et al. 2021; Reise et al. 2019). Originally designed to assess wood stock and 151 production, National Forest Inventories provide nationwide data on forest ecosystems that 152 allow, notably, to assess criteria and indicators of sustainable forest management and 153 ecosystem services (Simons et al. 2021). At the European level, data from National Forest 154 Inventories have been aggregated and published every five years for more than three 155 decades (Forest Europe 2020) and provide a set of indicators to inform sustainable forest 156 management.

In this process, the Sustainable Forest Management Criterion 4 is dedicated to the *"Maintenance, conservation and appropriate enhancement of biological diversity in forest ecosystems*". Composed of ten indicators, this criterion aims at helping decision makers to assess biodiversity-friendly sustainable forest management at the national and European scales. However, most of the indicators are indirect (structural) proxies of the forest ecosystem state (e.g. deadwood or forest fragmentation), whose link with biodiversity often

163 lacks strong scientific evidence (Gao et al. 2015; Paillet et al. 2018; Storch et al. 2023; Zeller 164 et al. 2022). Indeed, only two indicators involve other species than trees (namely 4.8 165 Threatened forest species, 4.10 Common forest bird species, Table 1) and to date, despite a 166 large corpus of individual studies and few syntheses (Zeller et al. 2023), no global and 167 systematic assessment of the correlations between multi-taxonomic biodiversity and Forest 168 Europe indicators has been attempted.

In this context, our aim was to provide an *ex post* validation of these indicators and propose improvements to the reporting process. For this, we: i) assessed the link between indicators and several - multi-taxonomic (i.e. taxonomic and functional groups) - *indicanda*; ii) identified the indicators that performed best and universally - i.e. for all groups; iii) defined, if possible, the most effective combination of indicators for forest biodiversity.

174 To address these aims, we first tested Forest Europe's biodiversity indicators against each 175 taxonomic and functional group. Second, we used a model selection approach to search for 176 the best indicators in terms of taxonomic and functional group. Third, we tested the effect of 177 a combination of indicators by searching for the best and most parsimonious combination of 178 biodiversity indicators. We used a multi-taxonomic database (Burrascano et al. 2023; 179 Trentanovi et al. 2023) combining species inventories and forest structure to analyze the 180 correlations between Forest Europe's biodiversity indicators with the biodiversity of six 181 groups (tracheophytes, epixylic and epiphytic bryophytes, birds, saproxylic beetles, 182 saproxylic non-lichenized fungi and epixylic and epiphytic lichenized fungi).

183

184 Materials and Methods

185 Database

We used the database gathered within the framework of the COST action "BOTTOMS-UP" (CA18207 – Biodiversity Of Temperate forest Taxa Orienting Management Sustainability by Unifying Perspectives). In a nutshell, this database comprises biodiversity samplings along with forest plot descriptions (Burrascano et al. 2023). It merges 34 different datasets from 12 European countries and more than 3500 sampling units. From this database, we extracted

the six most represented taxonomic and functional groups, namely tracheophytes, epixylic and epiphytic bryophytes (hereafter bryophytes), birds, saproxylic beetles, saproxylic nonlichenized fungi (hereafter fungi) and epixylic and epiphytic lichenized fungi (hereafter lichens).



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Figure 1: Spatial distribution of the sampling sites. Gray areas are covered by forests with a
tree cover greater than 40% according to Kempeneers et al. (2011). The size of the dot is
proportional to the number of sampling units.

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200 While tracheophytes and birds were inventoried without any specific selection of the guilds 201 targeted, only epiphytic and epixylic bryophytes and lichens were included (sampled on 202 living trees and deadwood), and saproxylic fungi and beetles (dependent on deadwood)203 substrates or on other organisms inhabiting deadwood).

204 Since the database is the result of the merging of different research projects, sampling 205 designs and protocols varied across datasets (Burrascano et al. 2023; Trentanovi et al. 206 2023). Therefore, we standardized species richness (number of species per sampling plot) 207 by dividing it by the asymptotic gamma richness at the site level, with site representing a 208 homogeneous sampling area with a maximum size of a few thousand hectares. We derived 209 site gamma diversity through plot-based rarefaction curves using the R package iNEXT 210 (Hsieh et al. 2016). Sites with less than 6 sampling units were discarded from the final 211 database since the estimation of the gamma richness was judged non-reliable. Exploratory 212 analyses revealed especially large deadwood volumes linked with small plots (nugget 213 effect). To avoid strong leveraging from these outliers (Zuur et al. 2010), we limited the 214 maximum deadwood volume per sampling plot to 500 m3/ha, a value that corresponds to the 215 maximal volumes observed in primeval forest of Europe (e.g. Christensen et al. 2005). The 216 final data distribution per taxonomic and functional groups is shown in Table 1 and Figure 1.

	I4.2.Regeneration			14	Total		
Таха	Coppice	Natural	Planting	Plantation	Semi- natural	Unmanaged	
Tracheophytes	10	734	94	94	550	194	838
Bryophytes	3	353	66	66	198	158	422
Beetles	7	408	94	94	255	160	509
Birds	10	721	94	94	537	194	825
Fungi	8	502	94	94	347	163	604
Lichens	10	332	92	92	333	9	434

Table 1: Number of plots used to fit the models distributed per taxonomic and functional

218 groups and Regeneration origin or Naturalness following Forest Europe (2020).

219

220 Indicators' calculation

We used the State of Europe's Forests (Forest Europe 2020) definitions to calculate the corresponding metrics for the indicators included in the Criterion 4 (Table 2). Since it was necessary to "translate" the definition into calculable values to analyze the relationship between indicator values and *indicanda* (i.e. standardized species richness), we proceeded as follows:

- indicator 4.6 *Genetic resources* was excluded since it was not possible to translate
 this indicator at the plot or site levels;

228 indicators 4.1 Diversity of tree species, 4.4 Introduced tree species and 4.5 229 Deadwood were directly calculated from the plot measurements. Instead of 230 introduced tree species, we considered only invasive woody species that can have a 231 significant impact on forest biodiversity following Campagnaro et al. (2018), namely 232 Acer negundo, Ailanthus altissima, Prunus serotina, Quercus rubra and Robinia 233 pseudoacacia (Campagnaro et al. 2018). We limited to these species since there 234 was no regional reference list for introduced tree species and it would have been far 235 beyond the scope of this study to create such a list, e.g. Norway spruce (Abies alba) 236 is introduced in lowland France but native in mountains where it has also been 237 planted in some places. In addition, no reference list of invasive tree species in 238 Europe was available in the Forest Europe (2020) report;

239 indicators 4.2 Regeneration and 4.3 Naturalness were derived from declarative -240 metadata when merging the database (Burrascano et al. 2023). In the 4.3, forests 241 were considered 'undisturbed by man' when declared without intervention by the data 242 holder (i.e. 'abandoned', 'unmanaged', 'primeval'). Time since the last intervention 243 was not used here since it was not homogeneously declared. 'Semi-natural' refers to 244 forest plots that are under any type of forest management regime (from even-aged to 245 uneven-aged silvicultural systems), but with natural regeneration processes of trees. 246 'Plantation' forests overlap with forest plots where regeneration derives from 247 'Planting' activities;

indicators 4.8 *Threatened forest species* and 4.10 *Common forest bird species* were
 derived from biodiversity sampling and reference lists. For 4.8, we used the IUCN
 Red List and calculated the total number of threatened species (categories VU, EN
 and CR) per plot all groups together. For 4.10, we calculated the total number per
 plot of the 34 forest birds classified by the Pan-European Common Birds Monitoring
 Scheme (https://pecbms.info/trends-and-indicators/indicators/);

- indicator 4.7 *Forest fragmentation* was calculated following the protocol applied by
 European Commission et al. (2019) on the Corine Land Cover 2018 layer
 (<u>https://efi.int/knowledge/maps/forest</u>). The forest area density (FAD) at fixed
 observation scale was obtained calculating the percentage of forest pixels in a 500
 ha circular buffer, centered on each plot;
- indicator 4.9 *Protected forests* was calculated based on the map of protected areas
 in Europe provided by the European Environment Agency (Nationally Designated
 areas CDDA: https://www.eea.europa.eu/data-and-maps/data/nationally-
 designated-areas-national-cdda-17
 We extracted the proportion of Categories Ia, Ib
 and II according to the IUCN in a 500ha circular buffer around each plot. We then
 added these three values so that the total proportion may be higher than 100.

Biodiversity indicators	Definition (Forest Europe 2020)	Data source	Metrics
4.1. Diversity of tree species	Area of forest and other wooded land, classified by number of tree species occurring	Plot measurements	Number of tree species per sampling unit
4.2. Regeneration	Total forest area by stand origin and area of annual forest regeneration and expansion (natural regeneration, planting, coppicing)	Metadata	3 classes: Natural regeneration Planting Coppicing
4.3. Naturalness	Area of forest and other wooded land by class of naturalness (undisturbed by man, semi-natural, plantations)	Metadata	3 classes: Unmanaged Semi-natural Plantation
4.4. Introduced tree species	Area of forest and other wooded land dominated by introduced tree species ⇒ modified to invasive ligneous species (Campagnaro et al. 2018)	Plot measurements	Number of invasive tree species per sampling unit
4.5. Deadwood	Volume of standing deadwood and of lying deadwood on forest and other wooded land	Plot measurements	Deadwood volume per ha
4.6. Genetic resources	Area managed for conservation and utilization of forest tree genetic resources (in situ and ex situ genetic conservation) and area managed for seed production	Not assessed	
4.7. Forest fragmentation (FAD)	Area of continuous forest and of patches of forest separated by non-forest lands (European Commission et al. 2019)	European Forest Institute map of forest cover	Forest area density (FAD) within 500ha around the sampling unit
4.8. Threatened forest species	Number of threatened forest species, classified according to IUCN Red List categories, in relation to total number of forest species	IUCN Red List species list	Species richness of threatened species (categories VU, EN and CR)
4.9. Protected forests	Area of forest and other wooded land protected to conserve biodiversity, landscapes and specific natural elements, according to MCPFE categories	CDDA map of protected areas	Cumulative percentage of Ia, Ib and II categories areas
4.10. Common forest bird species	Occurrence of common breeding bird species related to forest ecosystems	PECBMS	Species richness of forest birds

266 Table 2: Forest Europe's biodiversity indicators definitions and data sources used to

267 calculate them in this paper.

268

269 Statistical analyses

270 We processed all analyses in R v.4.3.0 (R Core Team 2023). We used the standardized

271 species richness of each taxonomic and functional group as a response variable. For each

272 group, we fitted a single generalized linear mixed model with each indicator as an 273 explanatory variable (9 models + the null - intercept only - model) and site as a random 274 effect on the intercept to account for the fact that two plots from the same site were more 275 likely to be similar than two plots from different sites. We used the beta error distribution with 276 logit link since standardized richness was continuous, strictly positive with a maximum value 277 below one. We scaled continuous explanatory variables to improve the convergence and to 278 make the estimates of the models comparable. We used the glmmTMB function in the 279 glmmTMB package (Brooks et al. 2017). We compared the magnitudes of all single effects 280 in the models and their significance and represented this using a heatmap of the 281 standardized effects (package ggplot2, Wickham 2016). We tested differences between 282 estimates of categorical variables (e.g. 4.3.naturalness) using Tukey post-hoc test. To 283 search for potential non-linear relationships between indicators and indicanda, we also fitted 284 generalized additive mixed models (GAMMs) with indicators as smoothers (package mgcv, 285 Wood 2023). We found very few differences between GLMMs and GAMMs and preferred to 286 stick to the first (comparisons are presented in Appendix 1).

287 Finally, to assess the relative performance of all indicators, we first compared the Akaike 288 Information Criterion corrected for small samples (AICc, Burnham & Anderson 2002) of all 289 single models including a null (intercept only) model. Then, we searched for the best and 290 most parsimonious linear combination (no interaction) of indicators that represented 291 biodiversity based on AICc using the dredge function (package MuMIn, Barton 2023). In this 292 process, we discarded the indicator 4.2. Regeneration since it was collinear with 4.3. 293 Naturalness and a model containing both variables could not be fitted (see Appendix 2). 294 When two competing models had a difference in AICc less than 2 points, we chose the most 295 parsimonious one.

296

297 Results

298 Data and indicators' distribution

299 The most represented group in the database was tracheophytes (838 plots) followed by 300 birds (825 plots, Table 3, see Burrascano et al. 2023 for a full description of the biodiversity 301 dataset). All taxonomic and functional groups have been inventoried in at least 400 plots 302 (Table 1). The distribution in the classes Regeneration (4.3.) and Naturalness (4.4.) were, 303 however, strongly unbalanced (see also Table 1): the majority of sampling units were 304 associated with "natural regeneration", while "coppicing" - and "planting" to a lesser extent -305 were underrepresented; the majority of sampling units were within semi-natural forests, but 306 the distribution was more balanced than for regeneration types. For quantitative indicators, 307 the values taken were relatively balanced between groups (Table 3) and no strong 308 collinearity was observed (see Appendix 2 for an example on tracheophytes).

Taxonomic and functional groups	I4.1.tree.sp	l4.4.invasiv e	I4.5.deadwood	I4.7.fragmentatio n (FAD)	l4.8.threa t.sp	14.9.IUCN	l4.10.bird s
Tracheophytes	1.96(2.06)	0.01(0.08)	41.83(60.72)	0.85(0.16)	0.25(0.56)	14.92(33.66)	3.63(1.78)
	[0-11]	[0-1]	[0-444.89]	[0.13-1]	[0-3]	[0-142.4]	[1-11]
Bryophytes	1.91(1.84)	0(0.06)	39.15(55.86)	0.84(0.18)	0.2(0.5)	27.02(43.16)	4.41(2.63)
	[0-11]	[0-1]	[0-461.41]	[0.06-1]	[0-3]	[0-142.4]	[1-15]
Birds	2.49(2.18)	0.01(0.08)	42.03(64.56)	0.85(0.16)	0.29(0.6)	15.92(34.06)	4.14(2.38)
	[0-11]	[0-1]	[0-461.41]	[0.13-1]	[0-3]	[0-142,4]	[1-15]
Beetles	1.96(2.01)	0(0.07)	41.1(57.2)	0.83(0.17)	0.26(0.59)	16.63(35.16)	4.15(2.43)
	[0-11]	[0-1]	[0-444.89]	[0.08-1]	[0-5]	[0-142.4]	[1-12]
Fungi	1.47(1.95)	0.01(0.08)	39.73(57.98)	0.82(0,.8)	0.17(0.49)	14.55(34.22)	4.32(2.83)
	[0-11]	[0-1]	[0-461.41]	[0.08-1]	[0-3]	[0-100]	[1-15]
Lichens	1.93(1.85)	0(0.06)	39.28(56.21)	0.84(0.18)	0.21(0.53)	26.65(42.93)	4.43(2.62)
	[0-11]	[0-1]	[0-461.41]	[0.06-1]	[0-5]	[0-142.4]	[1-15]

309 Table 3: Summary table for the range of values (mean per sampling unit (sd) [min - max])

310 taken for each indicator distributed over the different subsets. FAD = Forest Area Density.

- 311 Correlations between indicators and indicanda
- 312 Several of the Forest Europe indicators had significant relationships with the standardized
- 313 species richness of one or more of the six taxonomic and functional groups (Figure 2, see
- 314 Appendix 3 for values of the estimates).



Figure 2: Heatmap representing the standardized estimates (slope) of standardized species richness obtained with generalized mixed models with Beta error distribution and logit link, using Forest Europe indicators as independent predictors. S.N = semi-natural, UNM = unmanaged. Intercept (reference) for 4.2 Regeneration is the "Coppice" class, and "Plantations" for 4.3 Naturalness (p<0.1, * p<0.05, ** p<0.01, *** p<0.001).



Figure 3: Relationships between standardized species richness of 6 taxonomic and functional groups and Forest Europe indicator 4.5. Total deadwood volume. Estimations are issued from generalized linear models with Beta error distribution and logit link. Plain line

325 represents the mean estimate, ribbons the 95% confidence interval.

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Deadwood (4.5, Figure 3) and Diversity of tree species (4.1) showed four positive and (marginally) significant but generally weak relationships with tracheophytes, bryophytes, fungi and lichens (Appendix 3). Other indicators showed at least three significant (or marginally significant) relationships, with varying magnitudes, i.e. Fragmentation (FAD) (4.7), 331 Protected areas (4.9, Figure 4) and Forest Birds (4.10). Surprisingly, the proportion of 332 protected areas around sampling plots had a negative effect on birds and no effect on 333 tracheophytes and fungi. Regarding Naturalness (4.3, Figure 5), unmanaged forest and 334 semi-natural forests showed higher levels of fungi, lichens and bryophytes than plantations. 335 However, the opposite was true for tracheophytes. Threatened species (4.8) showed very 336 few significant results (marginally positive for tracheophytes and bryophytes, positive for 337 birds) while Invasive species (4.4) only showed a marginally significant positive relationship 338 with tracheophytes. Plantation (4.2) showed a very strong negative relationship with lichens 339 only, and a marginally significant negative effect on tracheophytes, while there were no 340 differences between coppice and natural regeneration (Appendix 3).



Figure 4: Relationships between standardized species richness of 6 taxonomic and
functional groups and Forest Europe indicator 4.9 (Proportion of protected areas in a 500ha
buffer). Estimations are issued from generalized linear models with Beta error distribution.
Plain line represents the mean estimate, ribbons the 95% confidence interval.

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Figure 5: Relationships between standardized species richness of 6 taxonomic and functional groups and Forest Europe indicator 4.3. Naturalness: PLA = Plantations; S.N = Semi – Natural forests; UNM = Unmanaged forests. Estimations are issued from generalized linear models with Beta error distribution and logit link. Barplots represent mean estimates, error bars are the 95% confidence interval. Letters indicate significant differences per group based on Tukey post-hoc test.

354 Relative performance of the different indicators

Comparing the AICc (Table 4) for all single models revealed that the indicator based on forest Naturalness (4.3) best explained the scaled species richness for three out of six groups (tracheophytes, bryophytes, and lichens); Regeneration (4.2) was within 2 points of AICc for tracheophytes and lichens. For birds and beetles, the indicator based on Forest birds (4.10) was the best explanatory one, while for fungi, Deadwood (4.5) stood first. Also note that the null model was never the best one and more than 2 AICc points away from the best.

		Tracheophytes	Bryophytes	Birds	Beetles	Fungi	Lichens
Explanatory variables	K	Delta AICc	Delta AICc	Delta AICc	Delta AICc	Delta AICc	Delta AICc
Null	3	14.5	76.9	626.2	2.6	123.1	27.4
I4.1.tree.sp	4	13.3	59.9	627.1	4.2	107.1	18.4
14.2.regeneration	5	0.5	23.5	629.5	6.0	73.3	1.6
14.3.naturalness	5	0.0	0.0	629.7	2.7	33.9	0.0
14.4.invasive	4	13.0	77.8	627.3	4.1	125.1	29.3
I4.5.deadwood	4	14.4	52.3	624.2	4.6	0.0	19.3
I4.7.fragmentation (FAD)	4	4.4	70.8	610.9	1.9	121.7	29.3
I4.8.threat.sp	4	12.8	75.2	623.8	4.4	124.1	28.5
14.9.IUCN	4	16.4	64.0	622.1	2.5	124.4	21.6
I4.10.birds	4	16.3	78.7	0.0	0.0	115.4	29.4

Table 4: Differences of Akaike Information Criteria corrected for small samples (Delta AICc) for all generalized mixed models with Beta error distribution and logit link. K is the number of parameters in each model. The best model (delta AICc = 0) is in bold.

365

366 Combination of indicators

We used data dredging to find the best linear (i.e. without interaction) combination of indicators reflecting the variations in species richness of each group. Only beetles were represented by only one indicator (Forest Birds, 4.10.), all the other ones had a combination

of 3 to 4 indicators in their best models (Table 5). In terms of indicators, Invasive species (4.4.) was never selected. Conversely, and as observed before, Diversity of tree species (4.1.), Naturalness (4.3.), Deadwood (4.5.) were selected four times, followed by Forest Birds (4.10, 3 times). Finally, it is interesting to note that, except for Fragmentation (FAD, 4.7.), all indicators had positive effects, and with the remarkable exception of beetles, all multivariate models performed better than single ones (in terms of AICc, they had at least 2

Taxonomic and functional groups	Tracheop hytes	Bryophytes	Birds	Beetles	Fungi	Lichens
Intercept	-0.47	-1.28	0.11	-1.03	-2.14	-1.34
I4.1.tree.sp	0.11	0.17			0.12	0.12
I4.3.naturalness	+	+			+	+
I4.4.invasive						
I4.5.deadwood	0.06	0.09			0.17	0.07
I4.7.fragmentation (FAD)	-0.072		-0.07			
I4.8.threat.sp			0.05			
14.9.IUCN						
I4.10.birds			0.65	0.05	0.08	
Df	8	7	6	4	8	7
AICc	-1246.18	-384.61	-1678.81	-1123.87	-1743.14	-547.79
Delta AICc	0.61	1.42	0.19	0.96	1.23	1.69
AICc (best single variable models)	-1226.72	-375.82	-1648.93	-1123.87	-1671.48	-541.632

points less than single variable ones, Table 5).

Table 5: Scaled estimates of the models selected by data dredging combining all indicators from Forest Europe (without interaction). Estimates are issued from generalized linear mixed models with Beta error distribution and logit link. "+" indicates the presence of the factor in the model. The best model was the most parsimonious (lowest degrees of freedom - df) with the lowest Akaike Information Criterion corrected for small samples (AICc). "delta" indicates the AICc difference with the best model. logLik = logarithm of the likelihood.

383 Discussion

384 We analyzed the links between Forest Europe's biodiversity indicators and the biodiversity of 385 six taxonomic and functional groups on a unique multi taxonomic dataset covering most 386 forest types in Europe (Burrascano et al. 2023). We showed that these indicators perform 387 unequally: while some correctly describe the biodiversity of several groups, no one acted as 388 a universal indicator, questioning their strength to predict biodiversity in European forests. In 389 contrast, our results reinforced the approach based on multiple complementary indicators of 390 the same *indicandum*. This also raises the question of contextualization of these indicators, 391 to better assess state and trends of biodiversity across Europe, and opens perspectives for 392 improvement.

393

394 *Performance of current forest biodiversity indicators*

While evidence on the link between some indicators and the biodiversity they are supposed to indicate remains incomplete (Gao et al. 2015; Penone et al. 2019; Zeller et al. 2023), our results highlighted several habitat-species relationships and habitat preferences of different groups.

399 We first observed strong effects of Regeneration origin (4.2) and Naturalness (4.3) on the 400 biodiversity of the six groups. The biodiversity of bryophytes and fungi was higher in 401 unmanaged forests compared to semi natural forests, and, including lichens, higher than in 402 plantation forests (Figure 5). On the contrary, plantations had a marginal positive effect on 403 tracheophytes compared to semi natural and unmanaged forests, and naturalness had no 404 effects on birds and beetles. The negative effects of plantations, and the positive effects of 405 management abandonment or primeval forests have been thoroughly documented (e.g. 406 Chaudhary et al. 2016; Paillet et al. 2010). However, our results should be nuanced by the 407 fact that the distribution of the data in the different categories of regeneration and 408 naturalness were strongly unbalanced, with only few plots located in coppice-managed 409 forests and a vast majority in semi-natural forests (Table 1). That said, unmanaged forests 410 had a generally positively influence on biodiversity in our dataset, especially for deadwood

411 dependent species (epixylic bryophytes and lichens), while plantations had detrimental 412 effects on several other groups. Indeed, habitat provision and continuity are higher in semi-413 natural forests, not to speak of unmanaged forests, which allow the persistence of dispersal 414 limited species (e.g. Boch et al. 2013; Boch et al. 2021). The positive response of 415 tracheophytes in plantations may seem surprising, but probably reflects a higher share of 416 disturbance tolerant herbs in more disturbed sites, as shown in several individual studies 417 (Boch et al. 2013). In addition, the semi-natural category encompasses a wide range of 418 management types (Trentanovi et al. 2023), and unmanaged forests in our dataset have 419 mostly been recently abandoned, which may cause a decrease in the biodiversity of 420 tracheophytes during the first decades after abandonment (e.g. Paillet et al. 2010).

421

422 Deadwood volume, followed by diversity of tree species, were two indicators that often 423 correlated with scaled species richness, both in univariate and multivariate analyses. 424 Deadwood and diversity of tree species have complementary roles for biodiversity (Storch et 425 al. 2023). Deadwood provides a substrate and a resource for numerous species that depend 426 on it for part of their life cycle (Lassauce et al. 2011; Müller et al. 2019; Müller et al. 2015; 427 Parajuli & Markwith 2023), but also for facultative species (Graf et al. 2022). Indeed, the 428 richness of bryophytes, fungi and lichens, and to a lesser extent birds, correlated positively 429 to deadwood volume in our analyses. Surprisingly however, saproxylic beetles richness did 430 not correlate significantly to deadwood volume despite a weak positive effect. It is likely that, 431 for this group in particular, deadwood does not act as a universal indicator that could be 432 transposed in all situations (see also Zeller et al. 2023). For example, Lassauce et al. (2011) 433 showed that the correlation between richness of saproxylic beetles and deadwood volume 434 was higher in boreal than in temperate forests, a result that was however not confirmed by 435 Parajuli and Markwith (2023). Müller et al. (2015) showed compensation effects between 436 deadwood and temperature, supporting that the influence of deadwood on saproxylic beetle 437 richness declines with increasing temperature, both in geographical space, but also locally in 438 relation to canopy openness (e.g. Koch Widerberg et al. 2012). It is also possible that the

439 observed deadwood volumes (mean 40 m³/ha, Table 3) lead to relatively poor beetle 440 communities. However, in most cases, the mean deadwood values in our dataset were 441 above the average deadwood volume per European country (which was 11.5m3/ha on 442 average, but varied from 2.3 to 28m3/ha between countries, Forest Europe 2020). 443 Consequently, while the positive role of deadwood for biodiversity remains unchallenged, its 444 use as an indicator should probably be refined by a better integration of the context 445 (macroclimate, microclimate, position on the deadwood volume gradient) in the relation. It is 446 also true that deadwood quality (e.g. including size and decay stage Vítková et al. 2018) is 447 as much relevant as its quantity. Some saproxylic beetle species (e.g. the red-listed species 448 Rosalia alpina, Campanaro et al. 2017) develop in large logs, snags and standing dead trees 449 within canopy gaps. In most of the sampling units we analyzed, deadwood was represented 450 by relatively small fragments (mostly around 15-20 cm diameter) likely deriving from thinning 451 processes occurring in closed forests (Burrascano et al. 2023). Further, many rare taxa 452 depend on the occurrence of tree-related microhabitats (Larrieu et al. 2018), such as the 453 protected umbrella species Osmoderma eremita, which larvae develop in large cavities in 454 living trees (Dubois et al. 2009). Since several tree-related microhabitats are very scarce in 455 managed forests (Paillet et al. 2017), and not readily guantified as part of the deadwood pool 456 (Müller et al. 2014), their role is not currently taken into consideration.

457 Diversity of tree species also showed positive correlation with the biodiversity of several 458 taxonomic and functional groups (bryophytes, fungi, lichens and more marginally 459 tracheophytes). Different tree species provide different habitat conditions for epiphytic and 460 saproxylic species living in forests (e.g. Boch et al. 2021; Cavard et al. 2011; Leidinger et al. 461 2021). These conditions encompass direct biotic interactions, different chemical properties of 462 the bark and wood, decomposition rates as well as differential light interception (Gosselin et 463 al. 2017; Zeller et al. 2023). This in turn provides niche heterogeneity for several species and 464 thus enhances biodiversity and confirms the role of diversity of tree species as a biodiversity 465 indicator.

466

467 Among the other indicators, forest fragmentation showed negative effects on tracheophytes 468 and birds, and a positive effect on bryophytes, and marginally on beetles and fungi. It should 469 be noted that since it is based on forest area density (European Commission et al. 2019), it 470 represents forest cover rather than its discontinuity and configuration, but has the advantage 471 of being calculable and homogeneous across Europe. Besides, the landscapes surrounding 472 our sampling units presented a high forest cover (77% on average, in a 500 ha buffer, 473 Table 1). Thus, the observed correlations reflect that forest area density could have been 474 beneficial for forest interior species that are sensitive to edge effects (bryophytes or lichens), 475 but detrimental to non-forest, open habitat or light demanding species, such as 476 tracheophytes and open-habitat birds (but see Bełcik et al. 2020, who show an increase 477 before a plateau in taxonomic diversity with increasing forest area in agricultural 478 landscapes). The inferences on the effect of fragmentation on biodiversity, based on the 479 Single Large vs Several Small principle (Diamond 1975), support that large continuous 480 habitat should have a greater effect on biodiversity conservation, compared to several small 481 ones. However, recent multitaxa studies proved the disproportionately high value of small 482 patches, harboring richer assemblages, even when focusing only on protected species (Riva 483 & Fahrig 2022). Besides, it has been highlighted that biodiversity conservation in human-484 modified forest landscapes is better achieved maintaining at least 40% of forest cover, rather 485 than focusing on fragmentation and configuration (Arroyo-Rodríguez et al. 2020). Hence, the 486 role of fragmentation, as it is currently estimated, should be probably be reconsidered to take 487 into account both the amount of forest cover and the value of the patches.

488

Contrary to the previous indicators, threatened species and birds are based on direct biodiversity sampling (other than trees). These indicators directly describe the evolution of a small part of the biodiversity, but generally correlate poorly to the richness of other taxa and most of the results observed are linked to the fact that these indicators represent subsets of larger groups (evidently forest birds vs. birds). However, despite several pieces of evidence showing that congruence between taxa is generally small, especially in forests (Burrascano

et al. 2018; Westgate et al. 2017), forest birds positively correlated to beetles and fungi.
Probably these groups respond to the same favorable habitat conditions, but this was not
reflected in the multivariate analyses.

The proportion of IUCN-protected areas around the plots positively influenced the richness of bryophytes and lichens, highlighting the positive effects of protection for these groups, but surprisingly had a negative effect on birds (see Honkanen et al. 2010).

501 Finally, invasive tree species did not show any correlation with biodiversity, apart from a 502 marginal positive effect on tracheophytes supporting high local tracheophyte diversity to be 503 often an indicator of disturbance. Despite the negative effects of invasive species 504 introduction on native flora, fauna and mycota, this effect is rather limited in our dataset 505 since it contains very few invasive species in total (maximum one tree species per plot) 506 reflecting that no projects targeted the effects of plantations rich in, or dominated by, 507 introduced tree species on biodiversity. This limited gradient probably does not allow to show 508 significant effects on biodiversity. In addition, an indicator on introduced species rather than 509 invasive ones – as it is defined in Forest Europe's indicators – may perform better.

510

511 Limitations of Forest Europe indicators to predict biodiversity patterns and implications for 512 improvements

513 Our study is one of the first to test Forest Europe indicators against multitaxonomic data at 514 the European level. During the model selection process, the best single variable models 515 were more often involving naturalness (tracheophytes, bryophytes and lichens) followed by 516 forest birds (birds and beetles) and deadwood (fungi). However, the multivariate models did 517 perform better than single ones (except for beetles), and often the best ones combined 518 several indicators to reflect variations of the *indicandum*. Only beetles were best indicated by 519 forest birds, but with a low magnitude, and birds by three combined indicators (including 520 forest birds and threatened species). The magnitude of the effects in the multivariate models 521 were comparable to those of single variable models (Appendix 3) which confirmed that 522 indicators were not collinear.

We studied only total richness as a response variable, and evidently, the results may be different for other, more specialized groups, or other metrics of biodiversity (abundance, occurrence of individual species, functional diversity). Such approach remains to be tested (see e.g. Lelli et al. 2019) but was beyond the scope of the present study. In addition, some of the references we used were probably incomplete regarding some groups: e.g. almost no lichens are included in the list we used for red-listed species, but the proportion of threatened species at the national levels may be high.

530 We limited our approach to the strict definition of the indicators as used in Forest Europe, but 531 higher performance could probably be reached by at least two improvements. First, adding 532 context to the indicators could probably reveal that they need to be adapted locally 533 (Chiarucci et al. 2012; Honkanen et al. 2010). Examples of context could be: elevation 534 (mountain vs. lowlands), biome (Mediterranean, temperate, boreal) and European forest 535 types. Second, most of the metrics we used are abundance metrics (apart from tree species 536 richness) that quantify habitat available for species. However, indicators based on diversity 537 of resources (following the heterogeneity-diversity theory, Tews et al. 2004) could perform 538 better, e.g. in the case of deadwood and saproxylic beetles (Bouget et al. 2013), or tree-539 related microhabitats and birds and bats (Paillet et al. 2018). It would then be interesting to 540 assess the performance of other indicator metrics vs. the current ones in assessing forest 541 biodiversity.

542

543 Conclusions: towards new indicators definitions and better reporting

544 Many forest biodiversity indicators are proxies based on pre-existing data mostly issued from 545 National Forest Inventories (Tomppo et al. 2010). Despite recent progresses based on 546 international initiatives (namely the Essential Biodiversity Variables, GEOBON, IPBES), 547 monitoring the state and trends of forest biodiversity solely based on proxies is not 548 satisfactory: while proxies are generally easier to measure than species themselves, they 549 are prone to demographic effects such as extinction debts or colonization credits. In other 550 words, the presence of a given habitat - such as deadwood - does not guarantee the 551 presence of the species that depend on it (e.g. Paillet et al. 2018). In addition, the response 552 of biodiversity to a given indicator depends on the taxonomic or functional group studied 553 (Zeller et al. 2022), so no indicator may represent biodiversity overall. This was confirmed by 554 our analyses. We showed that, while some indicators performed correctly for several 555 taxonomic and functional groups in terms of significance and magnitude, we did not highlight 556 a predominant role of one indicator over the others.

557 Stevenson et al. (2021) claimed that indicators were often implemented without clear 558 considerations of their purposes and utility in terms of decision-making. We argue that, while 559 combinations of current Forest Europe indicators are useful to delineate general trends in 560 biodiversity, taking into account context and analyzing the performance of other - more 561 diversity-driven - metrics would help better reporting on biodiversity (e.g. Alterio et al. 2023; 562 Paillet et al. 2018). Such improvements would also be beneficial to the use of indicators 563 beyond general trends to include evaluating management and policy actions, decisions, or 564 set biodiversity targets (Stevenson et al. 2021). For example, we showed that larger 565 negative effects on biodiversity were observed in planted forests. This poses key challenges 566 for making the 3 billion trees planting promoted by the European Forest Strategy for 2030 a 567 beneficial action for biodiversity in forests (Sills et al. 2020), if high growing rate plantations 568 such as introduced Douglas fir (Pseudotsuga menziesii) or eucalyptus (Eucalyptus spp.) are 569 promoted against semi-natural forests. Conversely, promoting old-growth and unmanaged 570 forests, as well as restoration of monocultures and conversions towards semi-natural forests, 571 could have a positive effect on biodiversity. It is crucial to assess and balance these potential 572 effects with the use of current data available and biodiversity indicators before taking actions 573 or to modulate actions in favor of biodiversity against detrimental ones.

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757 Appendices

- 758 Appendix 1: Comparison of linear (GLMMs) vs non-linear (GAMMs) models analyses
- 759 between standardized species richness and several Forest Europe's biodiversity indicators.

















Appendix 2: Correlation between indicators of Forest Europe from the tracheophytes dataset. Figures on the right are Pearson correlation coefficients, biplots on the left and histograms on the diagonal represent the distribution of the data. Regeneration (4.2) and Naturalness (4.3) are categorical variables not represented here.



774

- 776 Appendix 3: Standardized richness estimates table for all single generalized mixed models
- with beta error distribution and logit link. Intercept for I4.2.Regeneration is "Coppice", and for
- 14.3.Naturalness "Plantation" (PLA). S.N = semi-natural forest, UNM = Unmanaged forests.
- se = standard error of the mean, pl = critical probability, (*) p<0.1, * p<0.05, ** p<0.01, ***
- 780 p<0.001.

Group	variable	estimate	se	р	
Tracheophytes	I4.1.tree.sp	0.0648	0.0359	0.0712	(*)
	I4.2.regeneration(natural regeneration)	0.1431	0.2606	0.5828	
	I4.2.regeneration(planting)	0.5186	0.2754	0.0597	(*)
	I4.3.naturalnessUNM	-0.3878	0.0889	<0.001	***
	I4.3.naturalnessS.N	-0.2073	0.1429	0.147	
	I4.4.invasive	0.0376	0.0193	0.0508	(*)
	I4.5.deadwood	0.0318	0.0219	0.1463	
	I4.7.fragmentation	-0.0898	0.0256	<0.001	***
	I4.8.threat.sp	0.0428	0.0221	0.0522	(*)
	14.9.IUCN	-0.0087	0.0416	0.8344	
	I4.10.birds	0.0122	0.0326	0.707	
Bryophytes	I4.1.tree.sp	0.2828	0.0633	<0.001	***
	I4.2.regeneration(natural regeneration)	0.4156	0.4546	0.3606	
	I4.2.regeneration(planting)	-0.5511	0.4716	0.2426	
	I4.3.naturalnessUNM	0.8606	0.128	<0.001	***
	I4.3.naturalnessS.N	1.5727	0.1745	<0.001	***
	I4.4.invasive	-0.0409	0.0388	0.2917	
	I4.5.deadwood	0.2097	0.04	<0.001	***
	I4.7.fragmentation	0.1384	0.0486	0.0044	**
	l4.8.threat.sp	0.0776	0.0402	0.0539	(*)
	14.9.IUCN	0.2268	0.0585	<0.001	***
	I4.10.birds	0.0235	0.0571	0.6806	
Birds	l4.1.tree.sp	0.0341	0.0323	0.2916	
	I4.2.regeneration(natural regeneration)	0.0996	0.2114	0.6375	
	I4.2.regeneration(planting)	0.0452	0.2249	0.8407	

Group	variable	estimate	se	р	
	I4.3.naturalnessUNM	0.0388	0.0785	0.6207	
	I4.3.naturalnessS.N	0.2011	0.1259	0.1104	
	l4.4.invasive	-0.0177	0.0186	0.3417	
	l4.5.deadwood	0.0402	0.0203	0.0469	*
	I4.7.fragmentation	-0.1018	0.0243	<0.001	***
	l4.8.threat.sp	0.043	0.0205	0.0365	*
	14.9.IUCN	-0.1031	0.041	0.012	*
	I4.10.birds	0.6521	0.0218	<0.001	***
Beetles	I4.1.tree.sp	0.0242	0.034	0.4762	
	I4.2.regeneration(natural regeneration)	0.1411	0.2247	0.53	
	I4.2.regeneration(planting)	0.1782	0.233	0.4444	
	I4.3.naturalnessUNM	-0.0338	0.0638	0.5965	
	I4.3.naturalnessS.N	-0.0749	0.109	0.4917	
	I4.4.invasive	-0.0135	0.0175	0.441	
	l4.5.deadwood	0.0045	0.0204	0.8238	
	I4.7.fragmentation	0.0435	0.0263	0.0983	(*)
	I4.8.threat.sp	0.0096	0.0196	0.6255	
	14.9.IUCN	-0.0526	0.0357	0.1409	
	I4.10.birds	0.05	0.023	0.0297	*
Fungi	l4.1.tree.sp	0.1476	0.0343	<0.001	***
	I4.2.regeneration(natural regeneration)	0.1607	0.2444	0.5109	
	I4.2.regeneration(planting)	-0.3632	0.2542	0.1531	
	I4.3.naturalnessUNM	0.5017	0.0691	<0.001	***
	I4.3.naturalnessS.N	0.9083	0.1434	<0.001	***
	I4.4.invasive	0.0035	0.0189	0.8526	
	I4.5.deadwood	0.2228	0.0181	<0.001	***
	I4.7.fragmentation	0.054	0.0292	0.0644	(*)
	I4.8.threat.sp	0.0226	0.022	0.3039	
	14.9.IUCN	0.0422	0.0466	0.366	
	I4.10.birds	0.1021	0.0324	0.0016	**
Lichens	I4.1.tree.sp	0.1714	0.0507	<0.001	***

eration) 0.0169 -0.5447	0.258	0.9479	
-0.5447			
	0.2736	0.0465	*
0.5513	0.1028	<0.001	***
0.7687	0.2439	0.0016	**
0.0092	0.0313	0.7691	
0.1139	0.0343	<0.001	***
0.0137	0.0442	0.7564	
0.0339	0.0348	0.329	
0.1778	0.061	0.0036	**
0.0067	0.0559	0.904	
	0.1139 0.0137 0.0339 0.1778 0.0067	0.11390.03430.01370.04420.03390.03480.17780.0610.00670.0559	0.1139 0.0343 <0.001