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1 **Regional variation in deadwood decay of 13 tree species: effects of climate, soil**
2 **and forest structure**

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34

35 **Abstract**

36 Deadwood provides important ecological and biogeochemical functions in forest ecosystems.
37 These functions rely on tree species-specific decay dynamics, but only few studies investigated
38 how exogenous parameters affected the decay of deadwood at the regional scale. Here, we
39 explore the influence of climate, soil traits and forest structure on decay dynamics and mass
40 loss of logs of 13 tree species in 29 plots across three regions in Germany. This long-term
41 experiment (BELongDead) comprises 1,066 logs of 9 angiosperms and 4 gymnosperms and
42 covers a decay time of almost 10 years. Mass loss of logs was relatively high for the temperate
43 climate with partly different half-lives ranging from 6 to 13 years (mean 8.9 ± 2.2 years).
44 Diffuse-porous angiosperms lost 70 ± 13 % of their initial mass, followed by gymnosperms (49
45 ± 16 %) and ring-porous angiosperms (46 ± 12 %) within the study period. We applied three
46 different mathematical models to fit mass loss in time and found that a linear model is most
47 appropriate for 10 tree species. A sigmoidal model best described the mass loss of *Fraxinus*,
48 *Quercus* and *Pinus* logs. A multivariate analysis revealed a significant effect of soil,
49 temperature and precipitation on mass loss at the plot level. Mass loss increased with decreasing
50 soil nutrient content, possibly as a result of enhanced bidirectional element translocations
51 between nutrient-poor soils and logs by fungal hyphae. Temperature had a positive effect on
52 mass loss, whereas increasing precipitation and soil moisture were negatively related to mass
53 loss. The region with warmer, moderately humid climate and unfavourable soil properties led
54 to overall higher mass loss (66 ± 4 %) whereas the other two regions were indifferent (both
55 ± 3 %). Forest structure, including canopy cover, share of coniferous trees and the stock of
56 deadwood in the vicinity of the logs explained only a small part of the variability in mass loss.
57 High variability within individual tree species suggests that other factors such as organismic
58 diversity and microbial activity have stronger impact on the decay process at the regional scale
59 than exogenous factors.

60 **Keywords**

61 Coarse woody debris, deadwood, mass loss, tree species, climate, forest soil

62

63 **Highlights**

- 64 • Mass loss of logs of 13 tree species over 9.8 years is best described by a linear or a
65 sigmoidal model
- 66 • Logs of diffuse-porous angiosperms decay faster than ring-porous angiosperms and
67 gymnosperms
- 68 • Logs of the same tree species exhibit large variability in mass loss across 29 forest sites
- 69 • Mass loss is highest in the region with higher temperature, dry and nutrient-poor soils
- 70 • Forest structure including deadwood stock in the neighborhood had little influence on
71 mass loss

72

73 **Introduction**

74 Deadwood is important for biodiversity (e.g., Chmura et al., 2016; Jonsson et al., 2005; Seibold
75 et al., 2015; Stokland et al., 2012; Ulyshen, 2018) and carbon (C) storage (Cornwell et al., 2009;
76 Pan et al., 2011; Seibold et al., 2021) in forest ecosystems. **These functions rely on the residence**
77 **time of deadwood influenced by size, exposition (lying or standing) and tree species-specific**
78 **wood traits such as chemical composition and wood density (Christensen et al., 2005;**
79 **Kulakowski et al., 2017).** Dead lying trunks (logs) have contact to soil and are subjected to
80 relatively rapid decay or mass loss as opposed to standing dead trees (snags) (Storaunet and
81 Rolstad, 2002). Microbial decomposition, fragmentation into particles of different sizes, and
82 leaching of dissolved compounds contribute to deadwood decay (Harmon et al., 1986).
83 Microbial decomposition is the prevailing process, but the contribution of leaching and
84 fragmentation can increase at later decay stages and vary among tree species and environmental
85 conditions (Arthur et al., 1993; Lasota et al., 2018). Fungi are typically the key decomposers of
86 deadwood in temperate and boreal regions (Bässler et al., 2010; Jonsson et al., 2005), but
87 archaea, bacteria and micro- to macrofauna also use deadwood ingredients or its conversion
88 products and fragments for energy gain and buildup of biomass (Ceja-Navarro et al., 2014; Moll
89 et al., 2018; Seibold et al., 2021; Ulyshen, 2016). Structure and activity of the decomposer
90 community are complex and adapt to the changed wood chemistry as decay progresses (van der
91 Wal et al., 2013). There is evidence of a relationship between organismic diversity and decay
92 rates (Kahl et al., 2017), however, it is still unclear how the diversity of deadwood-degrading
93 organisms controls the log decay of individual tree species.

94 The decay of logs is also influenced by chemical wood traits, and the proportion of sapwood
95 and heartwood (Kahl et al., 2017; Noll et al., 2016). High concentrations of phenolic extractives
96 (e.g., quercetin) reduce the activity of extracellular enzymes and thus the biochemical
97 breakdown of the major wood polymers lignin, cellulose and hemicelluloses (Valette et al.,

98 2017). In addition, the content and composition of lignin may affect the microbial
99 decomposition. Especially the lignin components of gymnosperms are more stable than those
100 of angiosperms due to an increased level of guaiacyl subunits (Müller-Hagedorn et al., 2003).
101 Low availability of macronutrients such as nitrogen and phosphorous may further limit the
102 decomposition of deadwood (Cheng et al., 2011). Taken together, the chemical composition of
103 deadwood is tree-species specific, and can thus lead to different microbial activity and decay
104 of logs among tree species.

105 Exogenous factors such as climate, openness of canopies and other location-specific parameters
106 also play a significant role in deadwood decay (Harmon et al., 2020; Seibold et al., 2021; Zanne
107 et al., 2022). Temperature is known as a major driver of microbial activity, which increases
108 with rising temperature, and thus, controls the residence time of deadwood in forests (Rinne-
109 Garmston et al., 2019; Wang et al., 2002). Wood moisture, which depends on precipitation and
110 evaporation, additionally influences microbial activity in deadwood (Green et al., 2022).
111 **However, the control of wood moisture is more complex than temperature (Seibold et al., 2021).**

112 Both, low water availability during dry periods as well as high water contents with
113 correspondingly oxygen (O₂) depletion have been shown to limit the decomposition of litter in
114 general (Paul and Clark, 1989). Gas diffusion is ~10,000 times smaller in water than in air, and
115 thus high water contents limit the O₂ availability, especially inside the logs. Moreover, the
116 potential of low O₂ availability is higher in later than in early decay stages because of increasing
117 deadwood porosity and water-holding capacity (Covey et al., 2016). The canopy structure may
118 have no direct effect on the decay process, but an indirect effect through its influence on
119 temperature, water availability and radiation (Thom et al., 2020). While these exogenous factors
120 have been shown to affect decomposition, they have been mostly investigated over large climate
121 gradients. The importance of regional differences in exogenous factors and whether log decay
122 of individual tree species responds differently to exogenous factors has rarely been investigated.

123 Another exogenous factor that could affect the decay of logs is the underlying soil. Low nutrient
124 concentrations in deadwood promote the expansion of fungal hyphae in underlying soil to meet
125 their nutritional requirements. Nutrient translocations from soils into deadwood have been
126 described for N and P (Boddy and Watkinson, 1995; Philpott et al., 2014), but other limited
127 nutrients could also be taken up by fungal hyphae from soils and redistributed into deadwood
128 (e.g. metals that are required by lignin and cellulose oxidizing enzymes: Fe, Mn, Cu). Although
129 the mechanism is known, it is still unclear whether log decay is controlled by nutrient levels in
130 forest soils.

131 Simple decay models are commonly used to describe the temporal course of deadwood decay.
132 Further aims of modeling are the estimation of the decay constant, the residence time or half-
133 life of deadwood in a forest. Mass loss or density loss over time are frequently used to estimate
134 decay parameters with linear, exponential, or sigmoidal models (Chambers et al., 2000;
135 Freschet et al., 2012b; Müller-Using and Bartsch, 2009). A linear decay is given at constant
136 mass loss or density loss during the entire decay time. One pool exponential decay assumes a
137 first-order kinetic and declining decay activity per unit mass with decreasing mass or density.
138 Sigmoidal models indicate either fast or slow decay during the initial decay stage and thereafter
139 a decrease or increase in the decay dynamic, respectively. Estimated half-life or residence time
140 of deadwood allow comparisons of these models. These parameters as well as decay dynamics
141 can improve our understanding of the function of deadwood for inhabiting organisms and for
142 C storage in forests. However, there is no universal decay model that can be applied to coarse
143 woody debris of different tree species. Decay dynamics and best model type can even differ for
144 logs of the same tree species (Fraver et al., 2013). Therefore, we applied different decay models
145 to assess reliable estimates of decay constants or residence times for different tree species.

146 In this study, we investigated deadwood decay of 13 tree species in the framework of the large
147 scale field experiment 'BELongDead' (Biodiversity Exploratory Deadwood Experiment, Kahl

148 et al., 2017). Logs of 13 tree species were placed in 29 forests in three regions in Germany. We
149 asked the following questions:

- 150 1. Which model provides the best fit for the mass loss of logs of different tree species?
- 151 2. How does log decay of 13 tree species differ between regions with similar climatic
152 conditions?
- 153 3. What is the influence of temperature, moisture, soil traits and forest structure on log
154 decay?

155 **2. Methods**

156 **2.1 Study areas**

157 The BELongDead (**B**iodiversity **E**xploratories **L**ong-term **D**eadwood) experiment is an
158 interdisciplinary study on the diversity and function of deadwood-inhabiting organisms during
159 the entire decay process of logs. It is part of the large-scale and long-term project 'Biodiversity
160 Exploratories' to study the relationships between biodiversity, land-use intensity and ecosystem
161 processes in three regions (exploratories) in Germany (Fischer et al., 2010). The exploratories
162 are located in the UNESCO Biosphere Reserve Schorfheide-Chorin in northeast Germany
163 (52°47'25"-53°13'26" N, 13°23'27"-14°08'53"E, 3-140 m a.s.l.), in the National Park Hainich
164 and surrounding Hainich-Dün area in central Germany (50°56'14"-51°22'43"N, 10°10'24"-
165 10°46'45"E, 285-550 m a.s.l.), and the UNESCO Biosphere Reserve Swabian Alb in southwest
166 Germany (48°20'28"-48°32'02"N, 9°10'49"-09°35'54"E, 460-860 m a.s.l.). With 700–1000 mm,
167 mean annual precipitation (MAP) is highest in the Swabian Alb, followed by Hainich-Dün
168 (500–800 mm) and Schorfheide-Chorin (520–580 mm). Mean annual air temperature (MAT)
169 increases from Swabian Alb (6.0–7.0 °C) to Hainich-Dün (6.5–8.0 °C) to Schorfheide-Chorin
170 (8.0–8.5 °C). Soils developed on Jurassic limestone (Swabian Alb), Muschelkalk limestone,
171 often covered by loess deposits (Hainich-Dün), and glacial till, often covered by glacio-fluvial
172 or aeolian sand (Schorfheide-Chorin) (Fischer et al., 2010). The soil types and chemical-
173 physical soil properties vary greatly within and between the exploratories (Minnich et al., 2021).
174

175 **2.2 Experimental setup**

176 The BELongDead experiment was established on a total of 30 plots (9 in Schorfheide-Chorin
177 (SCH), 9 in Swabian Alb (ALB), 12 in Hainich-Dün (HAI)) in 2009. A storm event caused the
178 loss of one experimental plot in Hainich. The plots included different forest types and
179 management intensities, i.e., unmanaged forests dominated by European beech (n=9), managed
180 even- and uneven-aged forests dominated by European beech (n=12), and intensively managed

181 even-aged Norway spruce and Scots pine plantations (n=8). Logs of 13 tree species were
182 exposed in each of the three subplots per plot (in total 1131 logs). Tree species included nine
183 angiosperms: *Acer* spp. (*A. pseudoplatanus*, *A. platanoides*), *Betula pendula*, *Carpinus betulus*,
184 *Fagus sylvatica*, *Fraxinus excelsior*, *Populus* spp. (*P. tremula* and hybrids), *Prunus avium*,
185 *Quercus* spp. (*Q. robur*, *Q. petraea*), *Tilia* spp. (*T. cordata*, *T. platyphyllos*) and four
186 gymnosperms: *Larix decidua*, *Picea abies*, *Pinus sylvestris*, and *Pseudotsuga menziesii*. All
187 logs were 4 ± 0.1 m long and had a mean diameter of 31 ± 5.9 cm (SD). Living trees were felled
188 in the winter of 2008/2009 in the Federal State of Thuringia (Germany) to obtain fresh logs of
189 similar quality for the experiment. After transportation to each subplot, logs of the 13 tree
190 species were placed in random order beside each other with a distance of ~ 1 m between logs in
191 March to April 2009.

192

193 **2.3 Density, mass and volume loss of logs**

194 Initial wood densities of logs were first measured during the set-up of the experiment in 2009,
195 but they were underestimated due to methodological inaccuracies and were thus, substituted by
196 new measurements in 2020. For new measurements, three mature trees of each species of
197 similar diameter as described above were harvested in the state forest of Thuringia in the winter
198 of 2019/2020. Two ~ 10 cm thick wood discs were taken from each log's lower and upper section
199 using a chainsaw. Two representative subsamples of 300-600 cm³ of each wood disc were cut
200 and dried at 60°C until constant mass. Wood density (g cm⁻³) without bark of one subsample
201 per wood disc was determined by water displacement method after applying a thin layer of
202 paraffin to the wood surface. According to Archimedes' principle, water displacement of
203 paraffin-treated wood samples was measured at room temperature in a water-filled beaker using
204 a scale. Wood mass (60°C) was corrected for residual water content by drying a second
205 subsample at 103°C until mass constancy. It was assumed that the mean wood density of each
206 tree species from 2020 is representative for the initial wood density of logs in 2009.

207 Deadwood densities of each log were determined by different methods in 2012, 2015 and 2018.
208 The sampling methods differed between the years due to ongoing experiments and the need to
209 adapt the sampling method to progressive wood decay. In 2012 and 2015, after removing small
210 pieces of bark, holes were drilled in radial direction to the log center using a Forstner drill bit
211 (4 cm in diameter) and a cordless power drill at a 60 cm distance from one end of the log (see
212 Kahl et al., 2017). The Forstner drill bit allowed the drilling of exact cylindrical holes in logs.
213 Wood shavings from the drill holes were quantitatively collected and dried at 60°C until
214 constant mass. Wood mass at 60°C was corrected to drying at 103°C assuming 2% residual
215 water content for all tree species. Wood density (g cm^{-3}) was calculated as quotient of dry wood
216 weight (g) and drill hole volume (cm^3). To calculate the log-specific mass or mass loss in 2012
217 and 2015, the decline in log volume since 2009 had to be taken into account (see below,
218 equations 1 and 2).

219 In 2018, wood density of each log was determined from ~10 cm thick wood discs, and if
220 existing, also wood fragments that could be assigned to a disc. The discs were taken with a
221 chainsaw at a 20-30 cm distance from one end of the log. The exact distance to the end of the
222 log as well as mean diameter and thickness of each disc were immediately measured in the
223 field. All discs and fragments were placed in paper bags and dried at 60°C in ventilated ovens
224 for 2-4 weeks. After weighing, a representative pie-shaped piece was cut from each disc,
225 weighed and dried at 103°C to constant mass to determine the residual water content. Dry mass
226 of an entire disc with the associated fragments was calculated using the residual water content
227 (on average 2%). Deadwood density was assessed using the dry mass (disc and fragments) and
228 the volume of the disc. The volume of deadwood fragments was ignored, since these were rarely
229 found and the mass was small in relation to the disk (<5%).

230 To estimate the volume loss of logs, we made use of initial log diameters that were measured
231 from each log at five positions at a distance of 1 m in 2009. Linear interpolation between two
232 positions, assuming a conical log shape, allowed the estimate of the initial diameter for each

233 section of the log. The exact position and thickness of the disc taken in 2018 was used to
234 calculate the initial volume of the respective disc in 2009. Based on the initial (2009) and final
235 volume (2018) of a disc and the decay time between 2009 and 2018 (9.83 yr), the volume loss
236 factor (VLF , dimensionless) at time t was calculated for each log as follows:

$$237 \quad VLF_t = 1 - \left(\frac{1 - V_r}{9.83} \times t \right) \quad (1)$$

238 where V_r (dimensionless) is the relative volume of a wood disc in 2018 (normalized to 1 in
239 2009), 9.83 (yr) is the decay time between 2009 and 2018, and t (yr) is the decay time since
240 application of logs in 2009. This approach assumes a linear decrease in log volume over time.
241 Deadwood mass (M_t) was calculated for different sampling times as follows:

$$242 \quad M_t = D_t \times V_{2009} \times VLF_t \quad (2)$$

243 where D_t (g cm^{-3}) is the measured density of deadwood in year t , V_{2009} (cm^3) is the initial volume
244 of the disc taken in 2018. Initial deadwood mass in 2009 was calculated using the mean wood
245 density determined in 2020 (see above). Deadwood mass reported here refers to dry mass
246 (103°C) and assumes representative estimates of D_t , V_{2009} and VLF_t for each log.

247

248 **2.4 Mass loss and decay models**

249 Mass loss of each log was calculated as the relative difference (%) between the deadwood mass
250 in 2009 and 2018 (9.83 years). For modeling log decay, initial log mass was normalized to 1
251 and the respective deadwood mass in 2012, 2015 and 2018 was related to 1.

252 The average relative decrease in log mass of each tree species across all subplots was estimated
253 using three models:

$$254 \quad \text{Exponential: } RM = b \times e^{-kt} \quad (3)$$

$$255 \quad \text{Linear: } RM = b - kt \quad (4)$$

$$256 \quad \text{Sigmoidal: } RM = \frac{1}{1 + e^{(c(\log(t) - \log(d)))}} \quad (5)$$

257 where RM represents the relative decrease in log mass as a function of time, k is the decay
258 constant (yr^{-1}), t is the time (yr), b is the initial mass (normalized to 1), c and d are parameter
259 estimates (dimensionless) of the sigmoidal model. Parameter estimates of the models were used
260 to predict the average time for loss of 50% (L50, half life) and 95% (L95, residence time) of
261 initial mass.

262

263 ***2.5 Environmental parameters***

264 Various environmental parameters were compiled using the database BExIS of the Biodiversity
265 Exploratories (<https://www.biodiversity-exploratories.de/en/public-data-bexis/>) to test their
266 effect on log decay. Mean annual temperature (MAT) at 10 cm height above the forest floor,
267 mean sum of annual precipitation (MAP) and soil water content (SWC) at 10 cm depth were
268 continuously recorded in each forest site since 2008 in 10 minute intervals and averaged for the
269 complete time series (2009 – 2018). Canopy cover (Canopy) was assessed via airborne LiDAR
270 survey during leaf-on conditions as the proportion of first returns above 2 meters height in 2009
271 (BExIS ID 30925, not publicly available, Gossner et al., 2021). The conifer share (Conifer) is
272 based on the basal area represented by conifers and was calculated based on forest inventories
273 conducted between 2009 and 2011 (BExIS ID 18270, Schall and Ammer, 2014). Canopy and
274 Conifer are described as fractions between 0 and 1. For aboveground deadwood stocks (CWD,
275 coarse woody debris without logs of the BELongDead experiment), the volume of stumps,
276 downed and standing deadwood > 25 cm in diameter were determined on an area of 100 m x
277 100 m during a deadwood inventory in 2012 (BExIS ID 24546, Schall et al., 2019). Chemical
278 and microbial soil properties (Fe, Mn, P, ergosterol concentrations and pH) of the top mineral
279 soil from 0-7 cm depth and at a distance of 8 m from the logs were determined in one subplot
280 of each forest site (Minnich et al., 2021). Soil data were available in BExIS under ID 25726 and
281 ID 25727.

282

283 **2.6 Statistical Analysis**

284 Prior to the analyses, we dropped all logs with incomplete time series (1066 logs remaining).

285 To answer our first research question, we fitted linear, exponential, and negative sigmoidal
286 models (see 2.4) to find the best-performing decay model for each tree species separately. The
287 best models were determined by comparing Bayesian Information Criteria (BIC, best model
288 indicated by lowest BIC). Random slope effects were included on plot level for each tree species
289 to account for the nested design. Additionally, we used the *offset* function within R to set the
290 intercept to a fixed value of 1. For the mixed linear and exponential functions, we used the *lme4*
291 package (Bates et al., 2015), and for the sigmoidal functions, we used the *medrc* package (Ritz
292 et al., 2013). Using the coefficients of the best models, we estimated the model-specific decay
293 times (years) for 50% and 95% mass loss of each tree species. To address our second research
294 question, we performed an ANOVA to test for differences in relative mass loss after 9.83 years
295 among tree species, regions and their interactions. For this analysis, mass loss was arcsine
296 transformed to achieve normality of the residuals. Furthermore, for this analysis, we added the
297 subplots nested within plots as a random effect to account for the experimental design.
298 Differences between tree species were determined via Post-hoc tests (Tukey) using the
299 *emmeans* (Lenth, 2022) and *multcomp* (Hothorn et al., 2008) packages in R.

300 To address our third research question, environmental parameters (see 2.5) were used to analyze
301 differences in mass loss of logs among the exploratories, forest types, and forest stand
302 characteristics at the plot level and identify the most relevant abiotic drivers of deadwood decay.
303 We used a principal component analysis (PCA, built-in *prcomp* function in R) to overview the
304 relationships of 12 environmental variables and mass loss (Table S1, Supplement). To
305 determine the effects of environmental variables, we checked the contribution and correlation
306 of each variable to the most important principal components. If necessary, environmental
307 parameters were transformed for the PCA to achieve normality of the residuals. We used box-

308 cox-transformation, except for the conifer share in forest stands, for which an arcsine
309 transformation was applied (Table S1). As *Acer* logs on plot HEW42 were missing, we filled
310 relative mass loss for this plot using the mean value for *Acer* in HAI (64%). All variables
311 (dependent and independent) were standardized (*decostand* function of the *vegan* package,
312 Oksanen et al., 2020) before the multivariate analyses. Additionally, we calculated linear
313 models of the relative mass loss of all tree species for each of the environmental parameters in
314 interaction with region. All model calculations were conducted in R version 4.0.3 (R Core
315 Team, 2020).

316 **Results**

317 *Decay models, half-life and residence time of logs*

318 According to the BIC values, the exponential model revealed the weakest fits to the mass loss
319 of all tree species (Table 1). In contrast, BIC values of the linear and sigmoidal models were
320 relatively close for most tree species. Mass loss of *Acer*, *Betula*, *Carpinus*, *Fagus*, *Populus*,
321 *Prunus*, *Tilia*, *Larix*, *Picea*, and *Pseudotsuga* were best explained by the linear model. Mass
322 loss of all other tree species was best fitted by a sigmoidal model, although the shape of the
323 sigmoidal fit differed among the tree species (Fig. 1). *Pinus* and *Quercus* showed higher mass
324 loss during the initial decay phase and slowed down as decay progressed. In contrast, mass loss
325 of *Fraxinus* was relatively low in the initial phase and accelerated in the following years.
326 Estimated half-lives (L50) of logs were mostly similar for the same tree species using the model
327 parameters of the linear, exponential and sigmoidal model (Table 1). However, the time
328 differences between the models sharply increased when estimating the 95% mass loss (L95) for
329 the 13 tree species. The linear model predicted the shortest (11 to 24 years) and the sigmoidal
330 model (28 to 623 years) the longest decay times for 95% mass loss. Larger deviations between
331 the two models were found primarily for tree species with low mass loss. The exponential model
332 yielded consistently lower decay constants (k) and longer residence times (L95) than the linear
333 model.

334

335 *Wood density and mass loss of tree species in different regions*

336 Mean initial wood densities varied between 0.44 and 0.70 g cm⁻³ among the nine angiosperms
337 and between 0.45 and 0.50 g cm⁻³ among the four gymnosperms (Table 2). In 2018, after 9.83
338 years of decay, angiosperms had mean wood densities of 0.17 - 0.46 g cm⁻³ and gymnosperms
339 of 0.21 and 0.35 g cm⁻³. Of all tree species, *Carpinus* had the lowest (0.17 g cm⁻³) and *Quercus*
340 the highest wood density (0.46 g cm⁻³). Accordingly, tree species-specific mass loss ranged
341 between 42 and 80%, with *Larix* as the slowest and *Carpinus* as the fastest decaying tree

342 species. Overall, mass loss significantly differed among tree species ($p < 0.001$). Mass loss of
343 *Picea* was not different to *Acer*, *Populus* and *Prunus*, but higher than of *Fraxinus*, *Quercus* and
344 the three other gymnosperms. Mean mass loss of diffuse-porous angiosperms was $>60\%$,
345 whereas ring-porous angiosperms and gymnosperms exhibited mass losses of $<50\%$, except
346 *Picea* (62%).

347 Mass loss of logs from the same tree species strongly differed within and across the three
348 regions, as indicated by large standard deviations (Table 2, Fig. 2). The ANOVA revealed a
349 significant effect ($p < 0.001$) of region on mass loss for all tree species (Fig. 2, Table S2). The
350 highest mass loss occurred at SCH, whereas the differences for individual tree species between
351 ALB and HAI were mostly small and not significant. It is noticeable that *Acer* was the only tree
352 species with lower mean mass loss in SCH than in HAI and ALB, although the differences were
353 not significant. Mass loss of *Populus* and *Quercus* were also very similar and not significantly
354 different among the three regions. The significant interaction ($p < 0.001$) between region and
355 tree species indicated inconsistent patterns in mass loss for the 13 tree species between HAI
356 and ALB (Table S2).

357

358 *Effects of exogenous factors on mass loss*

359 The PCA showed the scattering of the 29 plots in the three regions, indicating large differences
360 in their characteristics both within and across the regions (Fig. 3). Particularly large differences
361 emerged between SCH and HAI or ALB, while few plots from ALB and HAI clustered. The
362 first two PCA axes explained 69.4 % of the variation in climate, soil parameters and forest stand
363 characteristics across all tree species and regions. The first PCA axis explained 50.7% of the
364 variability of exogenous parameters and was significantly correlated with mass loss ($R^2 = 0.52$,
365 $p < 0.001$), where the exploratory power decreased in the order soil Fe, Mn, SWC, P, pH, C:N,
366 MAT, MAP and ergosterol (Table S3 & Table S4). Thus, soil parameters had a stronger effect
367 on mass loss than MAT and MAP. PCA axis 2 explained 18.7% of the exogenous parameter

368 variation and was mainly driven by conifer share and canopy cover. Mass loss also correlated
369 significantly with PCA axis 2 ($R^2 = 0.13$, $p=0.04$). The plot-specific volume of CWD in the
370 neighborhood of logs had no discernible influence on mass loss (Table S3). As the interaction
371 between axes 1 and 2 was not significant, plot-specific mass loss responded similarly to the
372 exogenous factors (Table S4). Individual PCA analyses and linear models indicated that mass
373 loss of only nine tree species (*Betula*, *Carpinus*, *Fagus*, *Fraxinus*, *Prunus*, *Tilia*, *Picea*, *Pinus*,
374 *Pseudotsuga*) was correlated with axis 1 (Table S5). Moreover, axis 2 was only correlated with
375 mass loss of three tree species (*Fagus*, *Fraxinus*, *Larix*).

376 We further tested the effect of all exogenous factors shown in Fig. 3 on the mean plot-specific
377 mass loss separately for each region. Region-specific Pearson correlations between mean mass
378 loss of all tree species and site parameters revealed no significant relationships, except for soil
379 Mn concentration in SCH (Fig. 4). The concentrations of soil Mn were smaller at SCH (0.24
380 mg g^{-1}) as compared to ALB (1.35 mg g^{-1}) and HAI (1.17 mg g^{-1}) and positively correlated
381 ($r=0.72$, $p=0.03$) with mass loss in SCH.

382

383 Discussion

384 After almost ten years, logs of diffuse-porous angiosperms showed the highest mass loss,
385 whereas smaller mass loss was observed for the ring-porous angiosperms *Fraxinus* and
386 *Quercus*, and the four gymnosperms. Average half-lives of 6-13 years of the 13 tree species
387 indicate relatively rapid mass loss for the climatic conditions in the study regions. We further
388 explored how plot-specific or regional environmental factors impact the mass loss of logs.
389 Average higher mass loss occurred in the relatively warm-dry and nutrient-poor SCH region
390 (66%) as compared to HAI (57%) and ALB (57%). Soil properties, MAT and MAP had
391 relatively strong effects on plot-specific mass loss, whereas the effect of forest structure was
392 small. The high variability in mass loss for individual tree species suggests that other factors,
393 such as differences between the decomposer communities and microclimatic conditions in the
394 logs, create large variability in mass loss within and among plots.

395

396 *Limitations of decay models and half-lives of logs*

397 Mass loss was best described by either the linear or the sigmoidal model for the 9.8-year decay
398 phase. The exponential model yielded poorer fits to the mass loss, which could be related, at
399 least for some tree species, to the fact that the decay process was yet not completed and that
400 only four sampling dates were available for modeling. Changes in the resource quality through
401 preferential degradation of sapwood and retention of heartwood, or the stronger fragmentation
402 with improved accessibility for deadwood decomposers could have caused a turning point in
403 the decay dynamics. Therefore, we cannot rule out that the exponential, sigmoidal or another
404 model is more suitable to describe the entire decay process than the linear model. While the
405 sigmoidal model best describes the mass loss of *Pinus*, *Fraxinus* and *Quercus*, estimates of
406 residence times ($L_{95} > 100$ years) are unlikely for temperate regions. The sigmoidal model
407 requires thus more advanced decay for the prediction of the residence time of logs in our
408 experiment.

409 Overall, the linear model has the lowest uncertainty for the future decay process, especially for
410 tree species that had lost more than two-thirds of their mass. In agreement with our study, the
411 decay of *Fagus* and *Picea* logs in unmanaged forests also showed linear dynamics although
412 those logs had a history as standing deadwood (snags) before falling to the forest floor (Krüger
413 et al., 2014). In contrast, mass loss of *Betula*, *Picea* and *Pinus* logs with snag-history followed
414 sigmoidal decay dynamics in boreal forests (Mäkinen et al., 2006). A comparison with other
415 studies (Aakala, 2010; Freschet et al., 2012a; Müller-Using and Bartsch, 2009; Petrillo et al.,
416 2016) shows that decay patterns and models can differ for the same tree species. A combination
417 of several factors such as climate, local conditions, log size, cause of tree death and eventual
418 history as snag may influence the decay dynamics and the half-lives.

419 Taken together, half-lives of 6-13 years across all tree species are relatively short compared to
420 other studies under similar climatic conditions. For instance, Müller-Using and Bartsch (2009)
421 reported half-lives for *Fagus* logs of 10 years. Herrmann et al. (2015) estimated half-lives of
422 13 years for *Fagus*, 21 years for *Picea* and 22 years for *Pinus* logs. Likewise, Freschet et al.
423 (2012b) also calculated longer half-lives for *Betula*, *Pinus* and *Populus* logs. Tree species
424 specific mass loss has been described as a function of log diameter (Herrmann et al., 2015),
425 temperature (Russell et al., 2015), decomposer community (Kahl et al., 2017; Rajala et al.,
426 2012; Seibold et al., 2021), cause and conditions since tree death (Mäkinen et al., 2006). We
427 assume that the short half-lives in our study are related to the use of freshly cut trunks and the
428 almost complete contact of logs with the forest floor. The open cut surfaces on logs could have
429 accelerated the colonization and establishment of decomposer communities in the initial decay
430 phase. Moreover, soil contact creates favorable microclimatic conditions for microorganisms
431 and allows the uptake of nutrients via fungal hyphae and rhizomorphs from soil (see below).
432 These experimental boundary conditions have possibly contributed to the rapid mass loss of
433 logs in our study.

434

435 **Tree species-specific effects on mass loss**

436 Mass loss significantly differed for some tree species, which is probably related to different
437 wood traits as previously reported by Kahl et al. (2017) for our logs. At a higher phylogenetic
438 level, the differences were more pronounced with higher mass loss for diffuse-porous
439 angiosperms than for ring-porous angiosperms or gymnosperms. There was one exception -
440 logs of the gymnosperm *Picea* decayed relatively fast and were not different to logs of *Acer*,
441 *Populus* and *Prunus*. The rapid decay of *Picea* logs could be due to a brown-rot fungus,
442 *Fomitopsis pinicola*, which is common and adapted to the degradation of *Picea* deadwood
443 (Leonhardt et al., 2019). The linear course of mass loss does yet not indicate a decrease in decay
444 of *Picea* logs. In contrast, *Quercus* and *Pinus* logs showed an accelerated mass loss in the first
445 years of the experiment, which can be attributed to the rapid decay of sapwood. As the sapwood
446 of these tree species disappears, the sigmoidal decay dynamics may reflect a slowdown of mass
447 loss of recalcitrant heartwood in subsequent years. An exception is *Fraxinus* as it is the only
448 tree species that exhibits delayed mass loss during the first 2-3 years and thereafter an increase
449 in mass loss. To the best of our knowledge, a delayed decay of *Fraxinus* logs has not yet been
450 reported from other studies. Similar to mass loss, leaching of dissolved organic carbon (DOC)
451 from *Fraxinus* logs was also very low during the first years in our experiment (Bantle et al.,
452 2014). Leaching could be an important process in the initial decay phase as it removes
453 potentially inhibitory compounds such as soluble phenolics and aldehydes (Sanz et al., 2012).
454 Phenolics and aldehydes can affect enzymatic activities by inhibition and competitive reactions
455 (Aloui et al., 2004), and thereby the growth and spreading of wood-degrading microorganisms
456 in logs. Of the angiosperms, *Fraxinus*, *Prunus*, and *Quercus* had the highest initial levels of
457 phenolics (Kahl et al., 2017) as well as the smallest mass loss, but only *Quercus* and *Prunus*
458 logs showed high DOC leaching losses (Bantle et al., 2014). Hence, the level of phenolics and
459 their removal by leaching are possible drivers for the differences in mass loss among the 13 tree
460 species.

461 *Carpinus* is the tree species with the highest initial wood density and the highest mass loss.
462 Some logs of this tree species were almost completely decomposed within 9.8 years. Mass loss
463 of all other diffuse-porous angiosperms ranged between 61 and 74% and were partly
464 statistically different, but we cannot explain the differences among these tree species by initial
465 wood-chemical traits. It is often assumed that deadwood of angiosperms decays faster than that
466 of gymnosperms (Weedon et al., 2009). However, ring-porous angiosperms (*Fraxinus*,
467 *Quercus*) show similar or even lower mass losses than certain gymnosperms in our study.
468 Further ring-porous angiosperms like *Robinia* are also known as slow-decaying tree species
469 (Vek et al., 2020). The slower decay of ring-porous angiosperms could be related to the high
470 concentration of antimicrobial substances and phenolic antioxidants (Sanz et al., 2012).

471

472 *Effect of exogenous factors and region on mass loss*

473 The correlation between PCA axis 1 and plot-specific mass loss highlights the importance of
474 soil parameters and climate for the decay of logs. Soil parameters are closely related and
475 collectively suggest that nutrient-poor soils strengthen the decay of logs. This is particularly
476 true for soils at SCH, which are characterized by low pH, high C:N ratios and low nutrient
477 levels (Minnich et al., 2021). We attribute the higher decay potential of logs on nutrient-poor
478 soils to increasing expansion and activity of wood-inhabiting fungi in the underlying soil. As
479 deadwood is extremely nutrient-poor, uptake of nutrients from soil by fungal hyphae is a
480 pathway to overcome nutrient deficiency in logs (Philpott et al., 2014; Wells and Boddy, 1995).
481 The supply with readily available carbohydrates in logs allows deadwood-degrading fungi to
482 spread their mycelium in organic horizons of the forest floor or even in mineral soil horizons.
483 Soils beneath logs in the BELongDead experiment exhibited higher ergosterol (biomarker for
484 fungal biomass) concentrations (Minnich et al., 2021) and a shift in the fungal diversity towards
485 more saprotrophic and less mycorrhizal fungi (D. Persoh, pers. comm.). These changes were
486 stronger at SCH than at ALB and HAI because low soil fertility at SCH requires greater effort

487 to acquire the same amount of nutrients. On the other hand, if the level of a specific nutrient is
488 too small in the soil, mass loss may decrease. This seems to be an exception in the log-soil
489 continuum, but the relationship between soil Mn concentration and mass loss at SCH suggests
490 a limitation of fungal deadwood decomposition on extremely Mn-poor soils. A high Mn
491 availability is essential for the induction and function of the key lignin-degrading enzyme, i.e.
492 Mn-oxidizing peroxidase (Hofrichter, 2002). Biotic interactions between logs and soils have
493 been demonstrated, but further studies are needed to assess the impact of soil traits on mass loss
494 of logs.

495 Relatively high MAT and low MAP (Table S1) also contributed to higher mass loss at SCH
496 while climatic differences between ALB and HAI did not affect mass loss. The influence of
497 temperature on deadwood decay is expressed by a Q_{10} value of 2.4 (Chambers et al., 2000),
498 indicating an increase in mass loss by 14% by a temperature rise of 1°C. Consequently, the
499 higher mass loss at SCH can be explained, in part, by the temperature differences of 0.85°C and
500 0.51°C to ALB and HAI, respectively. The temperature difference of 0.34°C between HAI and
501 ALB was apparently too small for a significant regional effect. For logs of *Fagus sylvatica*,
502 highest mass loss was found in regions with warm-dry climate followed by warm-humid and
503 cold-humid regions (Přívětivý et al., 2016). Thus, temperature in interaction with humidity is
504 an important regional factor for mass loss.

505 It is noteworthy that mass loss was negatively affected by increased precipitation or soil
506 moisture across all regions since litter decomposition is typically positively associated with
507 water availability in well-drained forests (Borken and Matzner, 2009). This inverse pattern is
508 possibly related to the high water holding capacity of logs, especially with low wood densities
509 in later decay stages. Maximum water holding capacities varied between 230 and 750% in logs
510 from the BELongDead experiment after ~12 years and were substantially higher in diffuse-
511 porous angiosperms than in ring-porous angiosperms and gymnosperms (Groß et al., 2022).
512 High water contents can severely limit the oxygen supply and lead to anoxic conditions and low

513 microbial decomposition in logs. In fact, recent work showed the release of methane from logs,
514 which can only be formed by methanogenesis under anoxic conditions (Covey et al., 2016).
515 Hence, low precipitation associated with favorable oxygen availability in logs could have
516 promoted the activity of wood-degrading organisms and mass loss more strongly at SCH than
517 in the other regions. This effect on mass loss may have been enhanced by the sandy texture and
518 low field capacity of soils at SCH. (Bond-Lamberty et al., 2002) also reported higher deadwood
519 respiration on dry sites with well-drained soils compared to sites with poorly drained soils. On
520 the other hand, it can be assumed that deadwood decomposition is temporally limited under
521 very dry conditions due to insufficient water availability for microorganisms. As MAP and
522 SWC are only proxies for water contents in logs, monitoring log moisture could help unravel
523 the differences in log decay across individual tree species. We assume that canopy cover and
524 conifer share are interrelated with MAT and MAP and see no direct link to the decay of logs.

525

526 *Limitations of the study*

527 In this experiment, we used non-local logs with potentially different properties and nutrient
528 contents than logs from the study sites. It remains open to what extent the mass loss of these
529 logs would differ from local logs. To study this effect, logs of local origin would have had to
530 be applied in parallel. However, a study showed very similar nutrient levels in stem wood of
531 the same tree species from forest sites in three countries (Hagen-Thorn et al., 2004). We assume
532 that nutrient uptake from external sources (e.g., soil, throughfall, N₂ via biological nitrogen
533 fixation) is thus more important for the decay process than small differences in the initial
534 nutrient content of logs.

535 Tracking the mass loss of a single log over many years is based on some assumptions and
536 requires the use of different methods. When calculating the mass loss, it was assumed that the
537 density and decay state of the deadwood sample is representative of a total log although the
538 decay state of the logs was mostly inhomogeneous. Thus, the large variability for the mass loss

539 of a specific tree species across the three regions is partly due to the sometimes non-
540 representative deadwood sample of a log. However, the uniform sampling procedure and the
541 large number of replicates per tree species (Table 2) should result in a robust estimate of the
542 mean mass loss. With the inclusion of volume loss of logs, which amounted to 12-29% after
543 almost 10 years, the calculation of mass loss became more accurate. We recommend a higher
544 resolution of the initial log diameters to achieve a more accurate estimate of volume loss.
545 Progressing decomposition and fragmentation of logs will require further adjustments of the
546 sampling method in the future.

547

548 **Conclusions**

549 Decay of logs was relatively fast with half-lives of 6 – 13 years for the 13 tree species estimated
550 by linear or sigmoidal models, indicating that logs under similar environmental conditions can
551 rapidly lose part of their function as habitat and carbon storage. The almost complete contact
552 of logs with soils seems to accelerate the decay process in our study. Mass loss is highly
553 variable, primarily determined by tree species and to lesser extent by regional factors. In
554 temperate regions, high temperatures, low-moderate precipitation, well-drained and nutrient-
555 poor soils could amplify the decay of logs. However, the function of soils, in particular the
556 availability of soil nutrients, for the decay process remains unclear. Other factors as the diversity
557 and activity of wood-degrading organisms could also contribute to the variability in mass loss
558 at the local to regional scale. Assessment of tree species-specific mass loss is crucial for the
559 management of deadwood in forests. Rapid mass loss requires more frequent replacements to
560 maintain the deadwood stock and related ecological functions.

561

562

563

564

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584

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843 **Table 1.** Parameter estimates and BIC from linear, exponential and sigmoidal models of log decay of 13 tree species after 9.8 years, and predicted
 844 time (years) for mass loss of 50% (L50) and 95% (L95) by each model. Bold values indicate lowest BIC.

Tree species	Linear				Exponential				Sigmoidal				
	k (yr ⁻¹)	BIC	L50 (yr)	L95 (yr)	k (yr ⁻¹)	BIC	L50 (yr)	L95 (yr)	c	d	BIC	L50 (yr)	L95 (yr)
<i>Acer</i>	-0.070	-581.3	7.2	13.7	-0.108	-81.3	6.4	27.7	1.32	6.62	-578.7	6.6	61.3
<i>Betula</i>	-0.076	-491.5	6.6	12.5	-0.132	152.7	5.3	22.7	1.33	5.76	-451.4	5.8	52.5
<i>Carpinus</i>	-0.083	-655.3	6.0	11.5	-0.162	291.6	4.3	18.6	1.94	5.40	-543.4	5.4	24.7
<i>Fagus</i>	-0.076	-681.7	6.6	12.6	-0.132	119.8	5.3	22.8	1.93	6.14	-631.9	6.1	28.3
<i>Fraxinus</i>	-0.039	-615.5	12.9	24.5	-0.052	-226.6	13.4	58.1	1.91	12.55	-625.9	12.6	58.6
<i>Populus</i>	-0.065	-557.2	7.7	14.6	-0.102	33.3	6.8	29.4	1.86	7.52	-501.0	7.5	36.8
<i>Prunus</i>	-0.061	-364.0	8.2	15.6	-0.092	-15.1	7.5	32.4	1.32	8.14	-303.5	8.1	75.7
<i>Quercus</i>	-0.055	-639.5	9.2	17.4	-0.073	-388.1	9.5	41.0	0.72	10.44	-767.2	10.4	623.1
<i>Tilia</i>	-0.069	-530.5	7.3	13.9	-0.114	144.3	6.1	26.4	1.45	7.14	-487.5	7.1	54.6
<i>Larix</i>	-0.043	-604.6	11.6	22.0	-0.058	-236.7	12.1	52.1	1.00	14.42	-597.6	14.4	274.0
<i>Picea</i>	-0.067	-470.9	7.5	14.3	-0.102	-61.8	6.8	29.3	0.69	8.46	-429.2	8.5	589.1
<i>Pinus</i>	-0.054	-564.6	9.3	17.6	-0.075	-198.9	9.2	39.9	0.84	12.24	-628.5	12.2	402.5
<i>Pseudotsuga</i>	-0.043	-643.6	11.6	22.0	-0.056	-350.3	12.4	53.5	1.24	12.74	-623.5	12.7	136.7

845

846 **Table 2:** Means (\pm SD) of wood density (WD) of 13 tree species in 2009 and 2018, volume loss
 847 factor in 2018 (VLF), number of logs, mass loss in 2018, and significant differences in mass
 848 loss among tree species based on ANOVA results. Different letters indicate significant
 849 differences at $\alpha=0.05$.

Tree species	WD 2009 (g cm ⁻³)	WD 2018 (g cm ⁻³)	VLF 2018 (cm ³ cm ⁻³)	Logs (n)	Mass loss (%)	Significance
<i>Acer</i>	0.63 (0.01)	0.26 (0.08)	0.85 (0.10)	82	64.7 (10.5)	cd
<i>Betula</i>	0.65 (0.04)	0.20 (0.09)	0.85 (0.10)	84	73.5 (11.5)	b
<i>Carpinus</i>	0.70 (0.04)	0.17 (0.08)	0.78 (0.11)	85	80.4 (9.8)	a
<i>Fagus</i>	0.68 (0.08)	0.20 (0.07)	0.87 (0.09)	83	73.8 (9.2)	b
<i>Fraxinus</i>	0.61 (0.03)	0.40 (0.10)	0.88 (0.08)	85	42.6 (14.5)	f
<i>Populus</i>	0.44 (0.02)	0.20 (0.07)	0.77 (0.12)	81	65.7 (10.7)	cd
<i>Prunus</i>	0.53 (0.03)	0.21 (0.07)	0.82 (0.10)	60	61.4 (13.7)	d
<i>Quercus</i>	0.64 (0.03)	0.46 (0.07)	0.71 (0.11)	84	49.2 (7.6)	e
<i>Tilia</i>	0.50 (0.04)	0.21 (0.08)	0.77 (0.10)	83	67.6 (13.8)	bc
<i>Larix</i>	0.48 (0.05)	0.35 (0.08)	0.80 (0.10)	85	42.4 (14.0)	f
<i>Picea</i>	0.45 (0.03)	0.21 (0.06)	0.82 (0.10)	84	61.5 (13.0)	d
<i>Pinus</i>	0.50 (0.02)	0.30 (0.08)	0.84 (0.11)	85	50.0 (15.4)	e
<i>Pseudotsuga</i>	0.49 (0.03)	0.32 (0.07)	0.87 (0.08)	85	42.7 (12.3)	f

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851

852 **Figure captions**

853 **Figure 1.** Relative decrease in log mass of 13 tree species between 2009 and 2018, sorted by
854 gymnosperms, diffuse-porous and ring-porous angiosperms. The linear or
855 sigmoidal model was applied according to the lowest BIC (see Table 2).

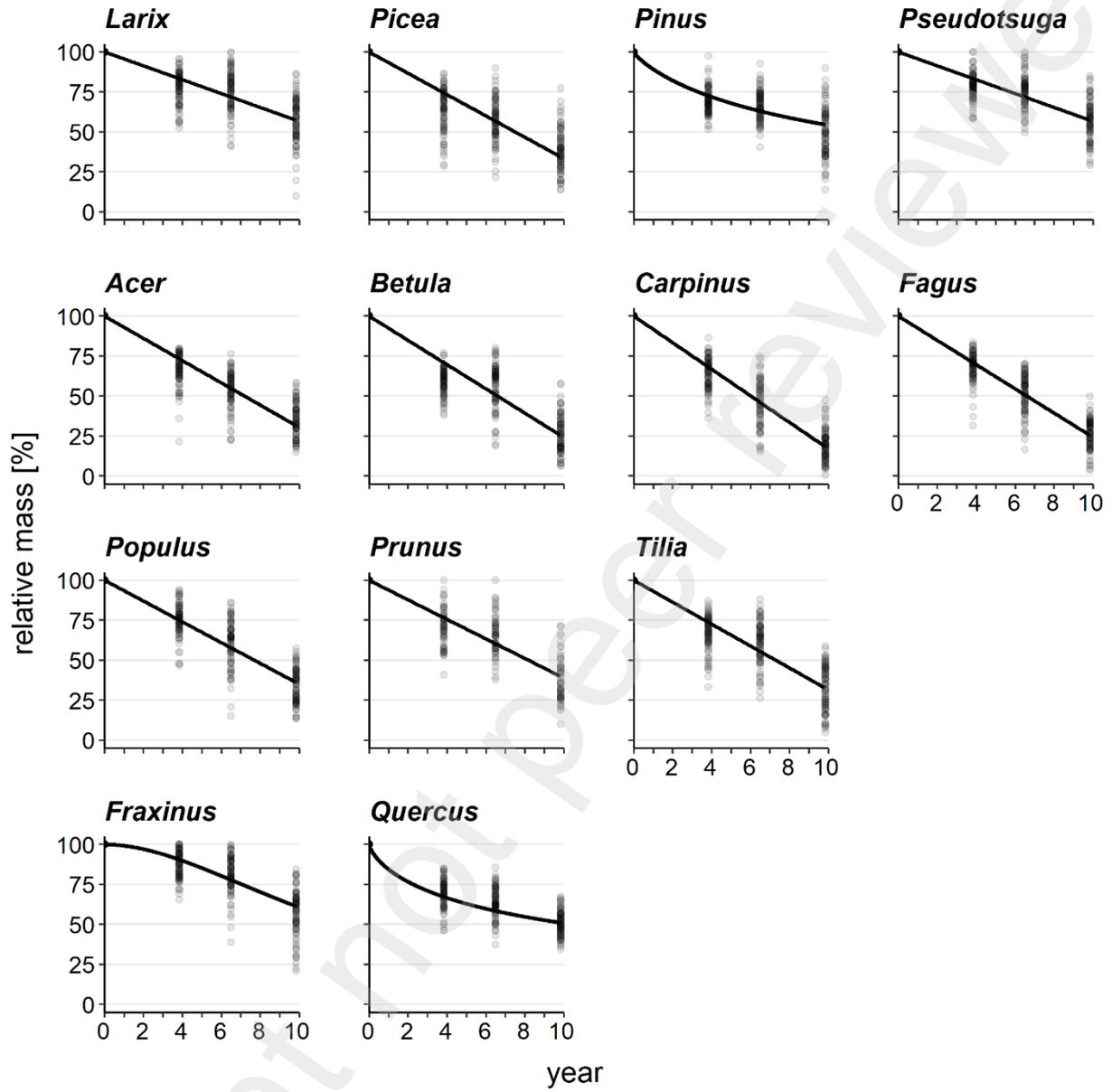
856 **Figure 2.** Boxplot (median with 25-75 percentile) with relative mass loss of 13 tree species in
857 three regions after 9.8 years.

858 **Figure 3.** Principle Component Analysis of mean mass loss of logs from 13 tree species,
859 climate parameters, soil properties, and forest stand characteristics across 29 sites in
860 three regions.

861 **Figure 4.** Relationship between mean mass loss of 13 tree species and soil Mn concentration
862 in the three regions ALB, HAI and SCH.

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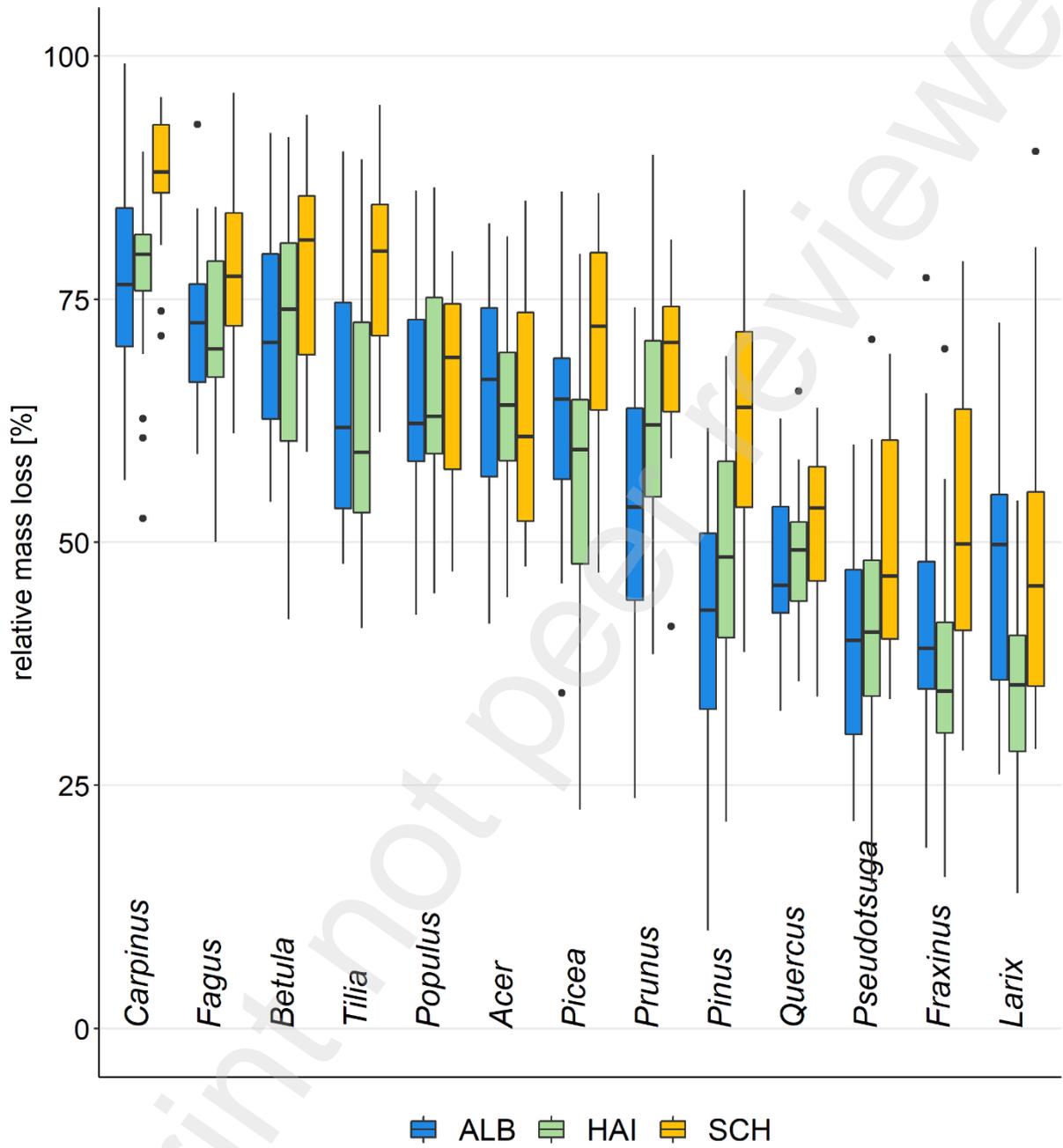
Figure 1



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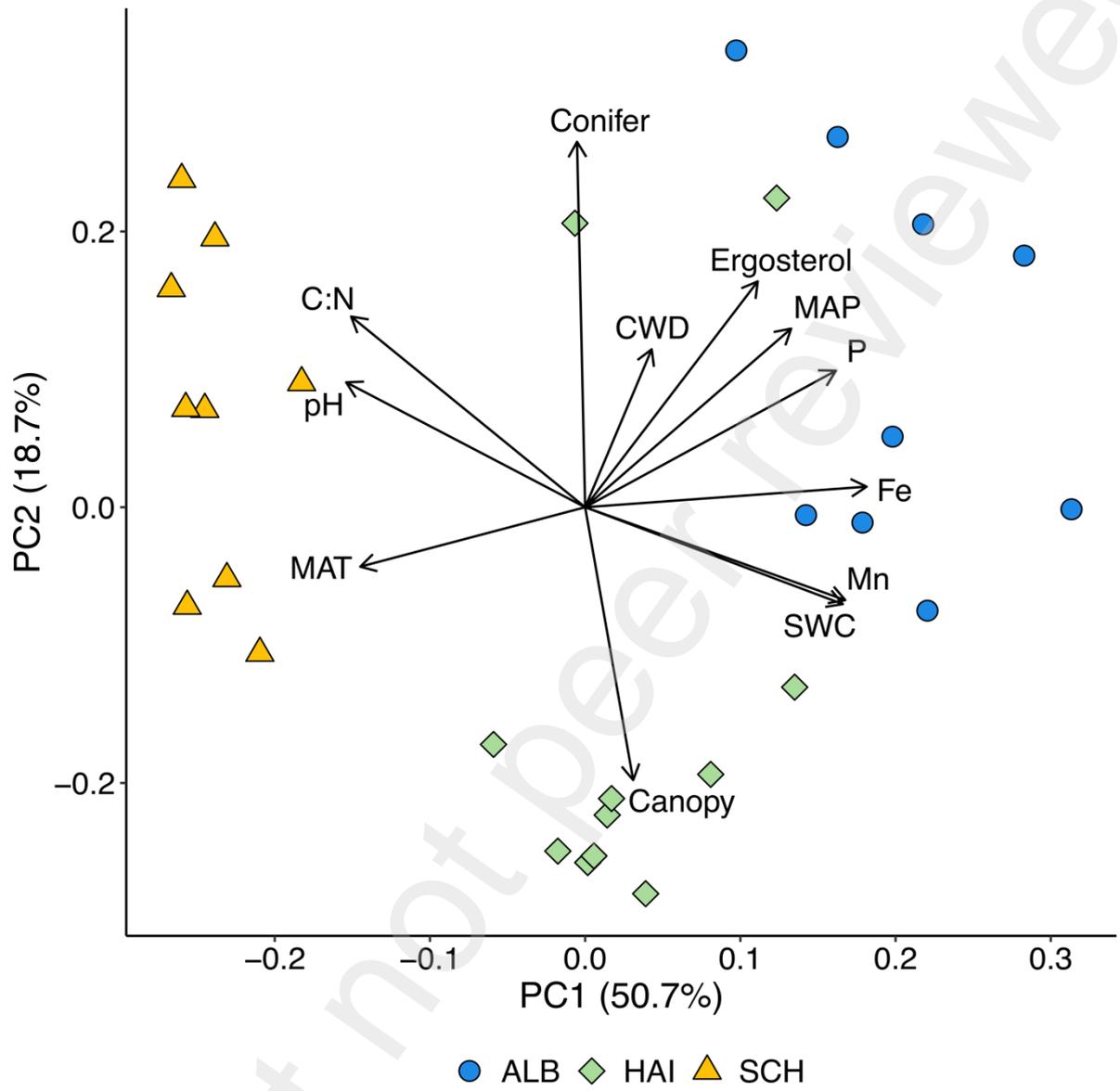
Figure 2



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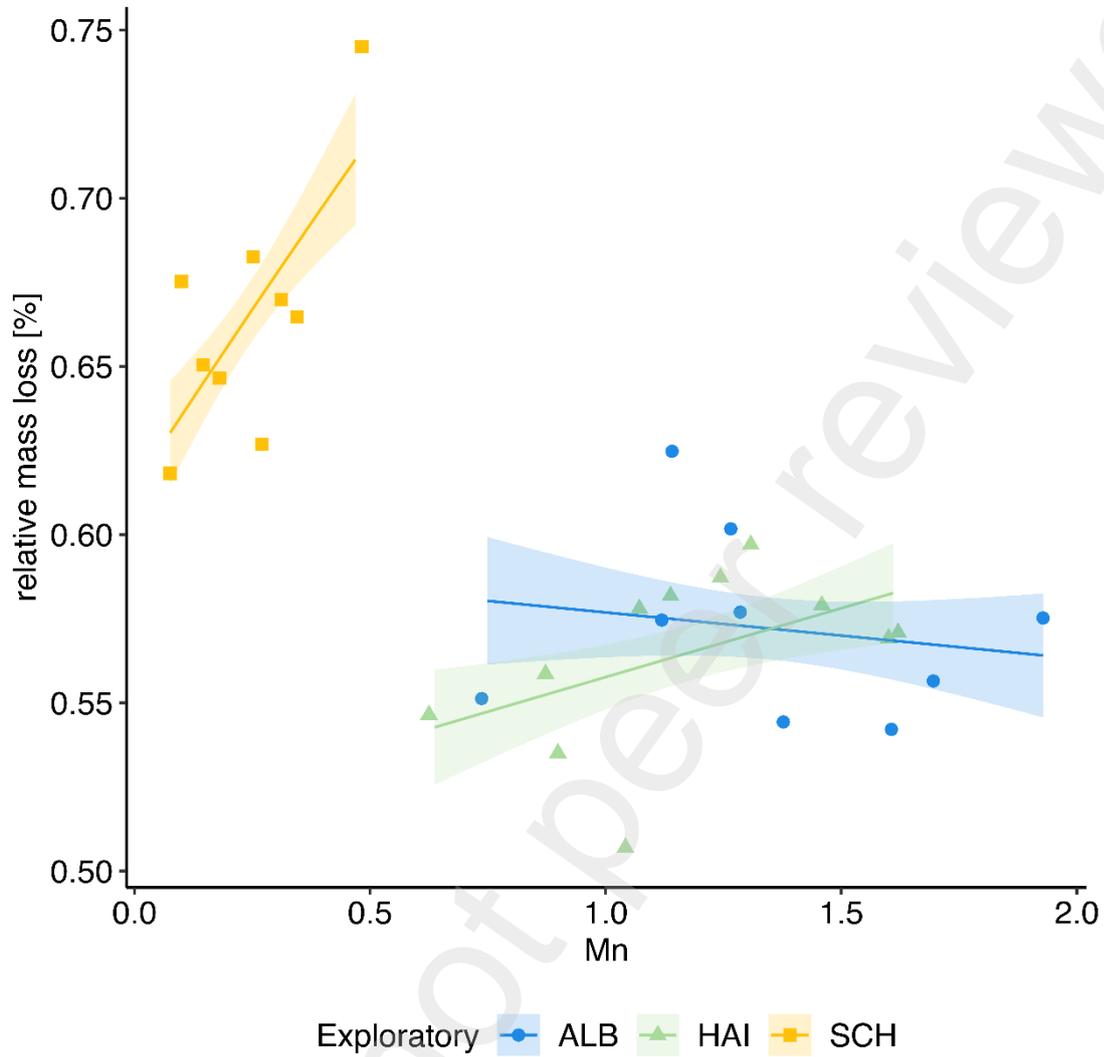
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Figure 3



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Figure 4



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Supplement

877 Table S1: Exogenous factors (mean \pm s.d.) for the three regions ALB, HAI and SCH used in
 878 the PCA. CWD (coarse woody debris) represents the deadwood stock without experimental
 879 logs. Transformation indicates that data were transformed to achieve normality of residuals.

Parameter	Description	ALB	HAI	SCH	Transformation	PCA
MAT	Mean annual air temperature at 10 cm [°C]	8.27 (0.32)	8.61 (0.41)	9.12 (0.29)	-	X
MAP	Mean annual precipitation [mm y ⁻¹]	942 (34)	590 (18)	601 (44)	-	X
SWC	Volumetric soil water content at 10 cm depth [%]	34.6 (4.2)	29.7 (5.6)	13.9 (2.6)	-	X
Canopy	LiDAR-based canopy cover [%]	90.3 (6.3)	90.8 (8.7)	86.1 (13.2)	Box-Cox Lambda: 1.99	X
Conifer	Share of conifers in plots [%]	42.7 (42.2)	16.1 (36.1)	39.5 (47.4)	Arcus-sinus	X
CWD	CWD stock [m ³ ha ⁻¹]	31.0 (17.2)	20.4 (11.9)	23.1 (10.3)	-	X
C:N	Ratio of soil organic C to total soil N	15.3 (1.0)	15.2 (1.0)	21.6 (2.6)	Box-Cox Lambda: -0.99	X
Ergosterol	Soil ergosterol concentration [μ g g ⁻¹]	19.8 (10.9)	8.5 (8.4)	6.3 (1.7)	Box-Cox Lambda: -0.14	X
P	Soil P concentration [mg g ⁻¹]	0.11 (0.05)	0.05 (0.02)	0.03 (0.01)	Box-Cox Lambda: 0.05	X
Fe	Soil Fe concentration [mg g ⁻¹]	37.0 (6.5)	24.8 (6.2)	5.4 (1.7)	Box-Cox Lambda: 0.64	X
Mn	Soil Mn concentration [mg g ⁻¹]	1.35 (0.35)	1.17 (0.31)	0.24 (0.13)	-	X
pH	Soil H ⁺ concentration [mol l ⁻¹]	2 \times 10 ⁻⁵ (4 \times 10 ⁻⁵)	10 ⁻⁵ (10 ⁻⁵)	10 ⁻⁴ (5 \times 10 ⁻⁵)	Box-Cox Lambda: 0.18	X

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882 Table S2: ANOVA results of mass loss by tree species and region.

Term	Sumsq	meansq	NumDF	Statistic	p
Tree species	19.65	1.64	12	121.69	<0.001
Region	0.96	0.48	2	35.55	<0.001
Interaction					
Tree species x Region	1.41	0.06	24	4.36	<0.001

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886 Table S3: Correlation and contribution of exogenous factors with first two axes of the PCA.

Variable	Correlation		Contribution (%)	
	Dim. 1	Dim. 2	Dim. 1	Dim. 2
MAT	-0.77	-0.14	9.77	0.87
MAP	0.71	0.42	8.20	7.81
SWC	0.88	-0.23	12.82	2.30
Canopy	0.16	-0.64	0.45	18.23
Conifer	-0.03	0.86	0.01	32.70
Fe	0.96	0.05	15.30	0.10
Mn	0.89	-0.22	13.07	2.11
pH	-0.82	0.29	11.04	3.82
CWD	0.23	0.37	0.86	6.11
C:N	-0.80	0.45	10.58	8.90
P	0.86	0.32	12.14	4.56
Ergosterol	0.59	0.53	5.76	12.49

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890 Table S4: Statistic results of mass loss by principal components 1 and 2.

Term	Coefficient	SE	NumDF	t value	p
PC 1	-0.016	0.003	1	-5.525	<0.001
PC 2	0.009	0.005	1	2.081	0.047

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893 Table S5: p values of linear models of arcsine transformed mass loss of 13 tree species and
894 principal components 1 and 2.

Tree species	PC 1	PC 2
<i>Acer</i>	0.82	0.47
<i>Betula</i>	0.03	0.12
<i>Carpinus</i>	<0.001	0.35
<i>Fagus</i>	0.03	0.02
<i>Fraxinus</i>	0.03	0.03
<i>Populus</i>	0.61	0.38
<i>Prunus</i>	0.001	0.07
<i>Quercus</i>	0.08	0.35
<i>Tilia</i>	<0.001	0.80
<i>Larix</i>	0.52	0.002
<i>Picea</i>	<0.001	0.22
<i>Pinus</i>	<0.001	0.80
<i>Pseudotsuga</i>	<0.001	0.07

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