

1 **Article type**

2 Letter

3

4 **Title**

5 Flow intermittence alters carbon processing in rivers through chemical diversification of leaf
6 litter

7

8 **Running Head**

9 Flow intermittence alters carbon processing

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11 **List of authors**

12 Rubén del Campo*^{†1,2,3}, Roland Corti^{†1,4}, Gabriel Singer^{1,3}

13 1.- Leibniz-Institute of Freshwater Ecology and Inland Fisheries (IGB), Berlin, Germany

14 2.- Department of Ecology and Hydrology, University of Murcia, Murcia, Spain

15 3.- Department of Ecology, University of Innsbruck, Innsbruck, Austria

16 4.- UR RiverLY, IRSTEA, Villeurbanne Cedex, France

17 †RdC and RC contributed equally to the development of the manuscript.

18 * Corresponding author at: Department of Ecology, University of Innsbruck

19 Technikerstrasse 25, A-6020 Innsbruck (Austria). ruben.del-campo@uibk.ac.at, telephone:

20 +43(0)512507-51737

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22 **Author contribution statement**

23 GS and RC conceived the study and designed the experiment; RC conducted the field and
24 laboratory experiments and collected the data; RC and RdC performed laboratory analyses. RdC
25 led data analysis with inputs of RC and GS; RdC led the writing of the manuscript; GS and RC
26 revised the manuscript critically and gave their final approval for publication. The authors
27 declare no conflict of interest.

28

29 **Data accessibility statement**

30 Whether the manuscript is accepted, all the data supporting our results will be uploaded to
31 Figshare.

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41 **Abstract**

42 The dry phase of intermittent rivers promotes the emergence of diverse terrestrial and aquatic
43 habitats where large amounts of leaf litter can accumulate. This environmental heterogeneity can
44 cause diverse chemical alterations in leaf litter by the co-occurrence of multiple physical and
45 biological degradation processes across these different habitats. After flow resumption, these
46 chemically diversified leaves are mixed and continue decomposition downstream in fully aquatic
47 conditions. We hypothesized that environmental heterogeneity during the dry phase can translate
48 into a chemical diversification of leaf litter, that may affect leaf litter decomposition in re-
49 established lotic conditions. Our laboratory treatments mimicking dry-phase habitats caused a
50 strong chemical diversification of leaf litter, which – upon combination in mixed litter bags –
51 accelerated its decomposition in a perennial river reach. Intermittent river reaches may act as
52 hotspots of organic matter diversification, with potential implications on C processing at river-
53 network scale.

54

55 **Keywords:**

56 Intermittent rivers, dry phase, chemical diversity, decomposition, river networks, biodiversity

57

58 **Scientific Significance Statement**

59 Intermittent rivers are among the most dynamic ecosystems worldwide. Their functioning is
60 regulated by the alternation of wet and dry phases. During drying, the cessation of surface flow
61 results in the emergence of a mosaic of diverse terrestrial and aquatic habitats (e.g. isolated
62 pools), where large amounts of riparian leaf litter can accumulate under contrasting
63 environmental conditions. After flow resumption, variously conditioned organic matter is mixed
64 and transported downstream; however, the implications for its final decomposition are unknown.
65 Here we demonstrate that spatial environmental heterogeneity during the dry phase of
66 intermittent rivers can result in a chemical diversification of the accumulated organic matter,
67 which in turn, can accelerate its decomposition further downstream, potentially affecting C
68 fluxes at river-network scale.

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76 **Introduction**

77 The fundamental role of hydrology in regulating C cycling in freshwater ecosystems (Raymond
78 et al. 2016) becomes particularly evident in intermittent rivers (Datry et al. 2018). There, the
79 alternation of dry and wet phases is thought to drive a pulsed processing of organic matter in
80 cycles of accumulation, transport, and decomposition (Larned et al. 2010; Datry et al. 2018).
81 Briefly, accumulation of organic matter – mainly riparian leaf litter - on riverbeds during the dry
82 phase (Sanpera-Calbet et al. 2016; Datry et al. 2018) is opposed to active processing during the
83 wet phase (Corti et al. 2011; Abril et al. 2016). In-between, the resumption of surface flow
84 mobilizes and transports material downstream (Corti & Datry 2012). This simplistic model has
85 been challenged, mainly because the exposure of leaf litter to various environmental conditions
86 during the dry phase (in this context also called “preconditioning”) can trigger chemical changes,
87 which ultimately affect leaf decomposition after flow resumption (Dieter et al. 2013; del Campo
88 et al. 2019; Mora-Gómez et al. 2019). For instance, the exposure of leaf litter to solar radiation
89 on dry riverbeds can increase its biodegradability due to photodegradation-induced loss of lignin
90 (Austin et al. 2016). In contrast, leaf litter accumulated in stagnant isolated pools can decrease in
91 biodegradability due to leaching of labile compounds and accumulation of phenols (Dieter et al.
92 2013). These are two examples from a great variety of terrestrial and aquatic habitats that emerge
93 during the fragmentation of water flow (e.g. wet and shaded remnant sediments, pools connected
94 to hyporheos, etc.) (Stanley et al. 1997; Datry et al. 2014). As accumulation and preconditioning
95 of leaf litter happen across this mosaic of aquatic-terrestrial habitats, chemical diversification of
96 leaf litter can occur (see Wickings et al. 2012, Fig. 1). The re-establishment of water flow then
97 triggers mixing of variously preconditioned leaf litter during downstream transport (see [here](#) an
98 example during a rewetting event in the Albarine river, France). Finally, upon retention,

99 decomposition proceeds in reassembled, chemically diversified litter packs in fully lotic
100 conditions.

101 Under aquatic conditions, decomposition is mainly controlled by leaf litter traits (Zhang
102 et al. 2019). The mixing of various species can have non-additive effects on decomposition
103 (Gessner et al. 2010), meaning that the decomposition rate of mixtures is either below or above
104 those expected from individual species` rates (Gartner & Cardon 2004). Negative effects of litter
105 diversity are associated to inhibition of decomposer activity by secondary metabolites like
106 polyphenols (Chomel et al. 2016). Positive effects are attributed to fungi-driven nutrient transfer
107 (Tonin et al. 2017) or nutritional complementarity among leaves with contrasting chemical
108 qualities (López-Rojo et al. 2020). Owing to the increased chance of obtaining essential
109 compounds, the functional diversity of leaf mixtures *per se* may accelerate decomposition
110 (Lecerf et al. 2011; Stoler et al. 2016).

111 Our understanding of chemical (bio)diversity effects on decomposition is still incomplete,
112 for instance with regard to natural mechanisms that could promote chemical diversity in
113 particulate organic matter beyond the diversity of leaf species. Indeed, the hydrological dynamics
114 of intermittent tributaries could create a powerful mechanism of chemical diversification of leaf
115 litter with unknown consequences for decomposition dynamics at river-network scale.

116 Intermittent rivers represent over half of the length of global river networks, and this fraction will
117 likely increase due to climate and global change (IPCC 2013); therefore, accounting for dry-
118 phase-associated diversity effects could be critical to achieve mechanistic understanding and
119 realistic modelling capacity for C fluxes at regional and global scale (Marcé et al. 2019).

120 Here we test the hypothesis that environmental heterogeneity occurring during the dry
121 phase of intermittent rivers can promote the chemical diversification of accumulated leaf litter

122 and thus, affect its decomposition in downstream rivers, once lotic conditions are re-established.
123 To this aim we first simulated the preconditioning of a single leaf litter species (*Alnus glutinosa*)
124 under various environmental conditions typically found during the dry phase of intermittent
125 rivers. Then, we measured the decomposition of leaf litter mixtures assembled using an
126 increasing number of preconditioning situations. We predict that the increase of chemical
127 diversity in mixtures of preconditioned leaves will accelerate decomposition in aquatic
128 conditions (Fig. 1).

129

130 **Materials and methods**

131 *Leaf litter preconditioning and preparation of mixtures*

132 We collected fresh leaves of *Alnus glutinosa* (alder) directly from several trees along the
133 Löcknitz river (Brandenburg, Germany) and let them air-dry for two weeks. Following
134 preconditioning through seven treatments (Table 1), we prepared fine- and coarse-mesh bags (0.5
135 and 8 mm, respectively, 15 x 15 cm size) containing leaves of single treatments (7 treatments x 4
136 replicates) and mixtures of leaves of increasing treatment richness in all possible combinations of
137 2, 4, and 6 treatments. This design resulted in 4 richness levels comprising a total of 91 bags (28
138 single-treatments + 21 2-treatment combinations + 35 4-treatment combinations + 7 6-treatment
139 combinations) for each mesh size. We filled each litterbag with 12 leaves. In mixtures, the 12
140 leaves were evenly partitioned across the component treatments. All leaves were scanned prior to
141 bag assembly to later measure treatment-specific leaf areas by digital image analysis (ImageJ,
142 <https://imagej.nih.gov/ij/>) and compute the exact contributions of component treatments on a dry
143 mass (DM) basis. For this, we established conversion factors of leaf area to DM (48h, 105 °C)
144 for each treatment from 20 leaves.

145

146 *Leaf litter chemical composition and calculation of chemical diversity*

147 Following preconditioning, sub-samples of all treatments were freeze-dried, ground using a ball
148 mill and analyzed for C- and N-content (Elementar vario EL C/N elemental analyzer, Germany),
149 other nutrients such as P, Ca, Mg and K by ICP-OES (Thermo Scientific, iCAP 6500, USA), and
150 macromolecular organic C moieties by Fourier-transform infrared spectroscopy (FTIR) (Duboc
151 et al. 2012; Liu et al. 2016; see detailed methods in Supplementary Information). We combined
152 information from FTIR peaks and elemental analysis to perform a single principal component
153 analyses (PCA) using z-standardized data. Average scores of each treatment on the first two
154 PCA axes served as a 2-dimensional proxy of chemical composition. To capture composition of
155 leaf mixtures we computed community-weighted means from the average of each involved
156 treatment's PCA scores weighted by its relative abundance in the mixture (see Stoler et al. 2016).
157 As a proxy of the chemical diversity of leaf litter, we computed Rao's quadratic entropy (RaoQ;
158 Stoler et al. 2016), a measure of functional diversity, using the package FD (Laliberte &
159 Legendre 2010) in R 3.2.1 (R Core Team 2015). RaoQ is the mean Euclidean distance among
160 treatments in the chemical space weighted by their relative abundance in the mixture. RaoQ was
161 considered 0 for litterbags containing single treatments.

162

163 *Aquatic decomposition experiment*

164 To measure aquatic decomposition of single treatments and mixtures, we incubated all litterbags
165 in the Löcknitz River (52°24'43.7"N, 13°49'33.6"E) for 23 days in August 2014. Löcknitz is a
166 forested, 3rd-order lowland river in the Elbe catchment (Germany). Litterbags were tied to iron
167 rods and fixed on the riverbed in four reaches of 50 m with running water and homogeneous

168 substrate, depth and flow conditions. During incubation the water temperature oscillated between
169 13 and 17 °C, average dissolved oxygen concentration was always above 6.5 mg L⁻¹,
170 conductivity and pH averaged 560 µS cm⁻¹ and 7.5, respectively.

171 After retrieving the litterbags at approximately 50% average mass loss, leaves were
172 washed individually in the laboratory with tap water above a 250 µm sieve to collect
173 invertebrates, which were preserved in 70% ethanol. Individuals were counted, identified to
174 family level and classified by guilds. The density of shredders was expressed as number of
175 individuals per DM of leaf litter. The leaves from each litterbag were dried (105 °C, 48h),
176 weighed, and leaf litter mass loss computed as the difference between initial and final litterbag
177 DM divided by initial DM.

178 From leaves in fine-mesh bags we cut a set of 12 discs with a cork borer (10 mm) to
179 measure fungal biomass as ergosterol according to Gessner et al. (2005) (Supplementary
180 Information). Values of ergosterol were expressed as µg g⁻¹ DM.

181

182 *Microbial respiration assay*

183 Parallel to the decomposition experiment, we measured oxygen consumption rates of
184 preconditioned leaves as a proxy for microbial respiration. We incubated 12 leaf discs by
185 mixture or single treatment in 250 mL sealed bottles filled with mineral water (Volvic) at room
186 temperature in a water bath. As microbial inoculum we used 10 mL of river water filtered by 0.7
187 µm pre-combusted glass fiber filters (Whatman GF/F, Maidstone, UK). Dissolved oxygen
188 concentrations were measured 13 times over 24 days with a needle-based micro-optode (PM-
189 PSt7 on a Microx 4 trace meter; PreSens, Germany). 10 bottles were filled with plain water as a

190 control. Oxygen consumption rates (day^{-1}) were computed as first order oxygen decay rates from
191 log-linear regression models.

192

193 *Data analysis*

194 To analyze the response of leaf litter decomposition to the increase of treatment richness in
195 mixtures we used generalized additive models for location, scale and shape (GAMLSS) (Rigby
196 & Stasinopoulos 2005). Models were built using the treatment richness (1, 2, 4, 6 treatments) as
197 explanatory variable and for the response variables mass loss, fungal biomass, shredder density
198 and microbial respiration. GAMLSS allow to model effects on the average values (μ) of the
199 response as well as its variance (σ). We also applied GAMLSS to test for the relationship
200 between treatment richness and chemical diversity in mixtures.

201 We estimated expected values of all response variables for each mixture using observed
202 values in single treatments, and compared those to observed values in mixtures. Expected values
203 were computed as the weighted average of observed values of component treatments on a DM
204 basis. Non-additive effects of mixing were considered synergistic when observed values were
205 significantly higher than expected based on paired Wilcoxon signed rank tests (Gartner &
206 Cardon 2004).

207 Finally, to analyze the influence of chemical diversity and chemical composition on the
208 decomposition of both single treatments and mixtures we used general linear models that
209 included RaoQ (chemical diversity), PC1 and PC2 (summary of the chemical composition traits)
210 as predictors. For each response variable, we built an initial model that included all main effects
211 and first-order interactions between RaoQ and each PCA axis and then selected a top set of most
212 parsimonious models with a multi-model inference approach (Grueber et al. 2011) using the R

213 package MuMIn (Bartón 2016). This top set kept all models with delta AICc < 2 to the best
214 model. Finally, using the natural method of model averaging we generated an average model
215 from the top set. This way we obtained a robust, weighted mean for each predictor coefficient
216 and its errors based on AIC weights (Grueber et al. 2011). All explanatory variables were z-
217 standardized to obtain scaled, comparable average predictor coefficients. We finally evaluated
218 the effect of chemical diversity and chemical composition on leaf litter decomposition by
219 comparing the absolute magnitude and direction of the averaged predictor coefficients and
220 checking whether their 95% confidence intervals spanned zero.

221

222 **Results and discussion**

223 *The dry phase of intermittent river as promotor of leaf litter chemical diversity*

224 The PCA based on the chemical traits of leaf litter clearly separated the various treatments (Fig.
225 2AB), evidencing a strong chemical diversification of leaf litter by the various treatments, that
226 were intended to mimic the environmental heterogeneity typically emerging in intermittent rivers
227 during drying. The observed chemical alterations of leaf litter agree with previous works (Dieter
228 et al. 2013; del Campo et al. 2019; Mora-Gómez et al. 2019). The greatest chemical
229 differentiation observed along PC1 (56 % of the total variance; Fig. 2B) was achieved between
230 distinct terrestrial and aquatic habitat conditions (Abril et al. 2016). Leaf litter in terrestrial-like
231 habitats (T1: leaves entering the riverbed shortly before flow resumption and T2: leaf litter
232 exposed to UVB radiation) retained labile carbohydrates, likely because of the limitation of
233 microbial activity by water scarcity (Abril et al. 2016). Conversely, leaves in aquatic habitats
234 with high temperature and nutrient concentration (T6 and T7) lost carbohydrates due to leaching
235 and microbial degradation (Dieter et al. 2013, Abril et al. 2016), while gained in nutrients and

236 phenols by microbial immobilization (Mora-Gómez et al. 2019). Separated from other treatments
237 along PC2, leaf litter immersed in stagnant pools (T5) increased in structural polysaccharides
238 such as cellulose, which might result from the (relative) loss of more soluble C compounds by
239 leaching due to limited microbial degradation in acidic and anoxic conditions (Dieter et al.
240 2013).

241

242 *Chemical diversity accelerates decomposition of leaf litter mixtures after flow resumption*

243 In litter mixtures, such chemically diversified leaf litter experienced accelerated decomposition
244 under fully aquatic conditions through synergistic effects on the activity of both microbial
245 decomposers and detritivores, in a similar way as reported for mixtures of riparian leaf litter
246 species (Gessner et al. 2010; Lecerf et al. 2011). Increasing preconditioning treatment richness in
247 leaf litter mixtures significantly increased mass loss in coarse- and fine-mesh bags, fungal
248 biomass and microbial respiration (Fig. 3A and Fig. 4), while this was not the case for expected
249 values computed from single treatments for any response variable (data only shown for mass loss
250 in Fig 3B). Shredder density did not increase significantly along the gradient of treatment
251 richness; but higher than expected shredder densities in leaf litter mixtures (Fig. SII) indicated a
252 positive effect of leaf litter diversity on the detritivore community as well.

253 Increasing treatment richness of leaf litter mixtures also caused a decrease of the
254 variability in fungal biomass and mass loss in fine-mesh bags (Fig. 4A and 4C). This often-
255 observed outcome of manipulating resource (or species) richness emerges by dampening of
256 extreme contributions in more complex mixtures (Dang et al. 2005; Lecerf et al. 2007). A
257 reduced variability in fungal-mediated decomposition with higher diversity of preconditioned
258 leaf litter translates to decreased spatial variability and increased stability of this ecosystem

259 process in downstream aquatic systems. Such a decrease in variance with treatment richness was
260 not found for shredder density or mass loss in coarse-mesh bags. This result may point to strong
261 influence of individual leaf litter types on the consumption of detritivores, which usually tend to
262 preferentially consume leaf litter species richer in labile C compounds or nutrients when present
263 in mixtures (Swan & Palmer 2006; López-Rojo et al. 2020).

264 Synergistic effects of leaf litter mixing on decomposition usually arise from facilitative
265 interactions among litter components with contrasting chemical composition. With our
266 experimental design, we cannot identify which precise mechanism drives the acceleration of leaf
267 litter decomposition by mixing; however, our results suggest chemical diversity as the main
268 factor stimulating decomposition. The increase in treatment richness in mixtures implied an
269 increase in chemical diversity (Fig. 2C). More importantly, chemical diversity was the main
270 predictor in averaged-models explaining mass loss in coarse-mesh bags, microbial respiration
271 and fungal biomass (Fig. 5). These results are in line with previous studies where chemical
272 diversity had a predominant influence on the decomposition of leaf litter mixtures (Lecerf et al.
273 2011; Stoler et al. 2016). We suggest that chemical diversity in mixtures of preconditioned
274 leaves enhanced the activity of microbial communities by facilitating the acquisition of essential
275 nutritional components for their growth and metabolism from multiple sources, such as nutrients,
276 labile C compounds like carbohydrates, or long-lasting resources like cellulose (Gessner et al.
277 2010).

278 Besides chemical diversity, the chemical composition of leaf litter strongly influenced
279 microbial decomposition of mixtures (Frainer et al. 2015; López-Rojo et al. 2020). The chemical
280 composition of leaf litter was the main predictor explaining mass loss in fine-mesh bags
281 (significantly positive effect of PC2) and microbial respiration (significantly negative effect of

282 PC1) (Fig. 5, Table S1) in averaged-models. The positive effect of PC2 scores on the mass loss
283 in fine-mesh bags may indicate a higher microbial activity associated with leaf litter rich in
284 cellulose (Talbot & Treseder 2012). On the other hand, the negative effect of PC1 on microbial
285 respiration could be due to either inhibition by lignin or phenolic compounds (Talbot & Treseder
286 2012; Chomel et al. 2016) and/or positive influence of carbohydrates (Stoler et al. 2016).

287

288 *Ecological implications at river-network scale*

289 Evidence highlighting the role of intermittent rivers in the processing of terrestrial organic matter
290 in drainage networks is continuously growing. Intermittent rivers are the most predominant lotic
291 ecosystem worldwide and they trigger hot moments of microbial processing of C during
292 rewetting events due to their capacity to accumulate large amounts of organic matter during the
293 dry phase (Datry et al. 2018; Marcé et al. 2019; Shumilova et al. 2019). Our results demonstrate
294 that, beyond the accumulation of organic matter, its chemical alteration and diversification
295 during the dry phase may have river-network scale implications. Our results suggest that flow re-
296 establishment in intermittent rivers triggers not only a pulse of organic matter, but a pulse of
297 chemical diversity, which is transported downstream across the river network, and consequently
298 may alter organic matter fluxes at regional, river-network scale. As mixing of variously
299 preconditioned leaf litter accelerates its decomposition, the length of organic matter transport
300 along the river network decreases. This means a spatial compression of organic matter
301 processing along the river continuum, which, in fact, counteracts the classical view of
302 intermittent rivers as pulsed bioreactors (Larned et al. 2010), where the organic-matter
303 processing length is considered to increase due to the little decomposition activity during the dry
304 phase and the far-reaching transport by flashy flow during rewetting events. Certainly, the

305 experimental character of our study precludes a strong assessment of implications. Future studies
306 will have to achieve this under natural conditions, also considering the influence of other factors
307 acting at larger spatial scale such as land use, climate or vegetation types. Here, we demonstrate
308 that environmental heterogeneity can promote chemical diversity, which in turn may accelerate
309 C processing in intermittent river networks. Our results reinforce the potential relevance of
310 intermittent rivers in global C cycling and the necessity (and difficulties) of integrating them in
311 larger scale modelling efforts.

312

313 **Acknowledgements**

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319 Ephemeral Streams, www.smires.eu). RC was funded by the IGB Fellowship Program in
320 Freshwater Science and the project FLUFLUX (ERC-STG 716196).

321

322 **Tables**

323

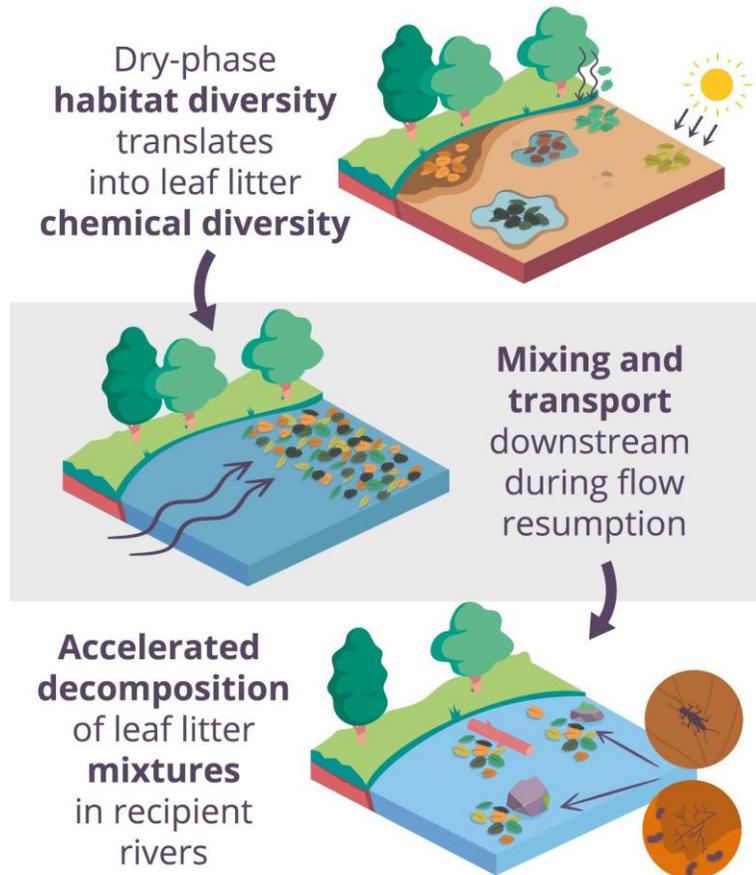
324 Table 1. Summary of the preconditioning treatments used in the study to mimic the terrestrial-aquatic habitat mosaic appearing in
 325 intermittent rivers during the dry phase.

Treatments	Riverbed habitat during the dry phase	Laboratory simulation	Physicochemical conditions in aquatic habitats
T1 Untreated	Vertical input of leaf litter shortly before flow resumption	Initially collected leaves, air-dried and kept at room temperature and in darkness.	
T2* UV	Dry riverbed exposed to intense solar irradiation	Irradiation for 12 h/day with a UV lamp (Cosmedico Arimed B6, Osram Biolux 965, Germany; with 31% UVB of total UV) at room temperature.	
T3 Moist	Shaded and humid riverbed habitats	Container with soil from the Löcknitz river floodplain moistened with 500 mL of tap water every 4 days and kept at room temperature.	
T4 Cold pool	Pool connected to hyporheic flow paths with cold and nutrient-poor water supporting limited algal growth	Aquarium filled with mineral water and stones with biofilm from the Löcknitz river. The aquarium was continuously illuminated, oxygenated by air-bubbling and kept at constant low temperature.	T = 15.3 °C DO = 9.45 mg L ⁻¹ pH = 7.94 Cond = 925 µS cm ⁻¹
T5 Anoxic pool	Anoxic, stagnant pool	Container filled with mineral water and 8 mg of Na ₂ SO ₃ per mg dissolved oxygen to create anoxic conditions, kept at room temperature and in darkness.	T = 24.6 °C DO = 0.15 mg L ⁻¹ pH = 5.5 Cond = 1650 µS cm ⁻¹
T6* Wet/dry	Habitats subjected to wet/dry cycles associated to rain events	Alternating T2 and T3 every 7 days.	
T7 Hot pool	Disconnected pool with warm and nutrient-rich water supporting algal growth	Same conditions as T4, except that the aquarium was kept at room temperature and a nutrient solution (0.6 g L ⁻¹ of NaNO ₃ and 0.3 g L ⁻¹ of KH ₂ PO ₄) was added.	T = 25.1 °C DO = 6.72 mg L ⁻¹ pH = 7.66 Cond = 800 µS cm ⁻¹

326 T: water temperature, DO: dissolved oxygen, Cond: water conductivity. *The duration of preconditioning treatments was 21 days except for T2
 327 and T6, which extended for 60 days, since terrestrial decomposition processes occur at a longer time scales than aquatic ones.

328 **Figure legends**

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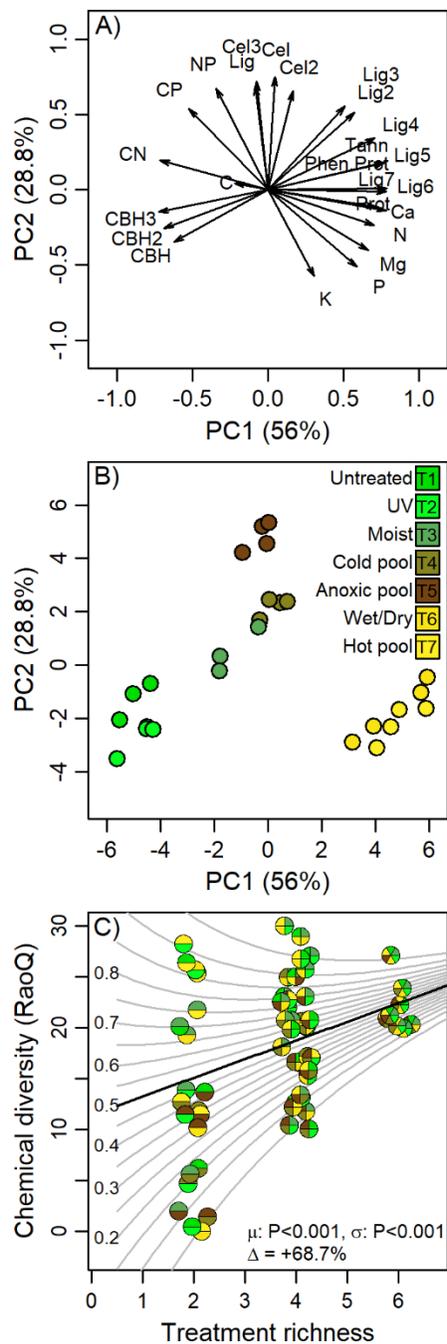
331 Figure 1. Conceptual figure showing the main hypothesis of the study. Environmental
332 heterogeneity promotes a chemical diversification of accumulated leaf litter during the dry phase.
333 After flow resumption, variously “preconditioned” leaves are mixed and transported
334 downstream. Upon retention, chemical diversity accelerates the decomposition of leaf litter
335 mixtures.

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341 Figure 2. PCA describing changes in the chemical composition of leaf litter due to
 342 preconditioning under different treatments (A and B). (A) Variable loadings defining the PCA
 343 space. (B) Distribution of the preconditioning treatments across the PCA space. In (B), the colors
 344 of treatments represent their positions in PCA-space; similar chemical compositions of two
 345 treatments (e.g. T6 - T7) translates to similar colors and vice versa (e.g. T1 - T7) – this allows

346 chemical interpretation of color in subsequent figures. (C) GAMLSS identified a significant
347 increase of the average (μ , black line) and a significant reduction of the variance (σ , grey
348 percentile lines) of chemical diversity (RaoQ) with increasing richness or preconditioning
349 treatments. The colors in the pie charts used as symbols for mixtures indicate chemical
350 composition as identified in (B).

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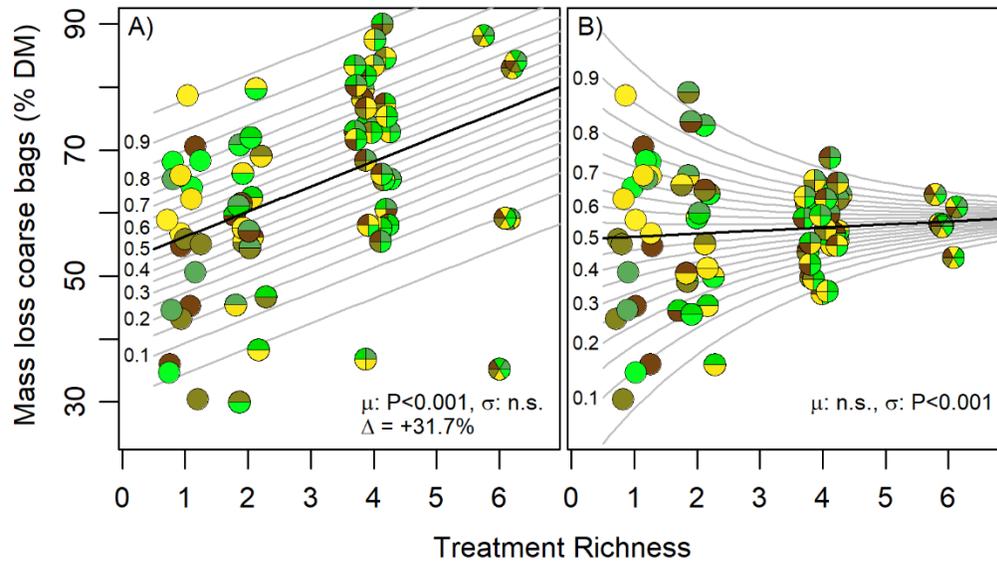
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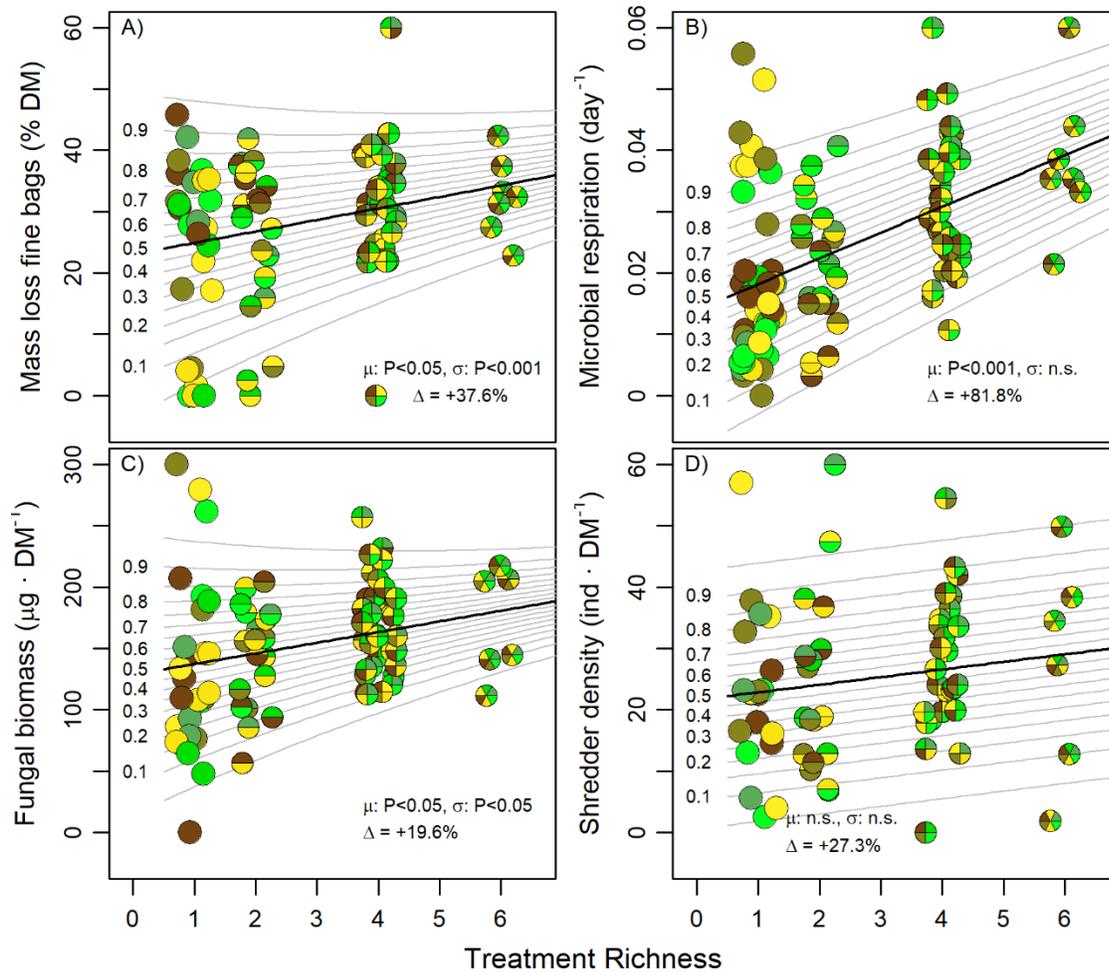
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370 Figure 3. Observed (A) and expected (B) values of mass loss in coarse-mesh bags in single
 371 treatments and mixtures along the treatment richness gradient. GAMLSS identified a significant
 372 increase of the mean (μ , black line) but no change in the variance (σ , grey percentile lines) of
 373 observed values of mass loss with increasing richness, while there was no change in the mean
 374 but a decrease in variance of the expected values. Colors indicate the identity of single treatments
 375 (simple dots) or the treatment composition of mixtures (pie charts); color codes in Fig. 2B.

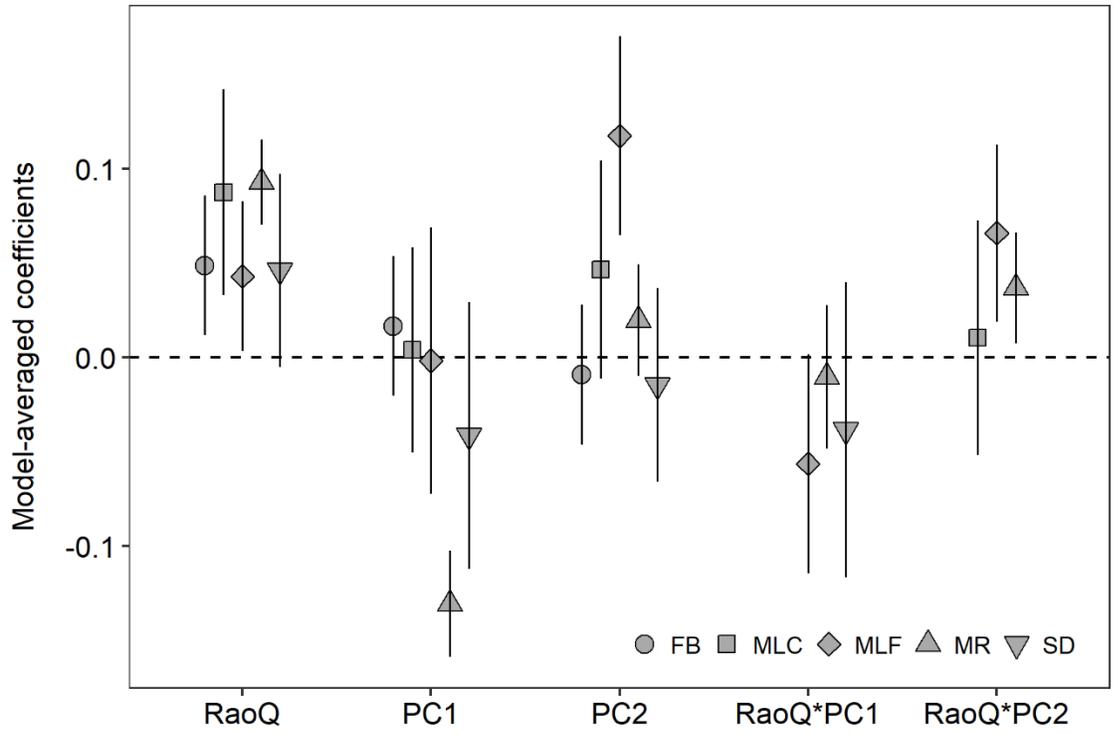
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 378 Figure 4. Observed values of mass loss in fine-mesh bags (A), microbial respiration (B), fungal
 379 biomass in leaf litter (C) and shredder density (D) in single treatments and mixtures along the
 380 treatment richness gradient. GAMLSS identified a significant increase of the mean (μ , black line)
 381 for all four variables with increasing richness, but a decrease in the variance (σ , grey lines) only
 382 for mass loss in fine mesh bags and fungal biomass. Colors indicate the identity of single
 383 treatments (simple dots) or the treatment composition of mixtures (pie charts); color codes in
 384 Fig. 2B.

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388 Figure 5. Model-averaged coefficients (mean \pm 95% CI) of predictors explaining the mass loss,
 389 microbial respiration, fungal biomass and shredder density of single treatments and mixtures.

390 Chemical diversity (estimated through RaoQ) was the most important predictor for mass loss in
 391 coarse-mesh bags, fungal biomass and shredder density, while chemical composition features
 392 (estimated through the average score of PC1 and PC2) were more important explaining mass loss
 393 in fine-mesh bags (PC2) and microbial respiration (PC1).

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