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1 **Pesticides and land cover heterogeneity affect functional group and taxonomic**
2 **diversity of arthropods in rice agroecosystems**

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20 **Abstract**

21 Biodiversity can be characterised across several dimensions, which are crucial for the
22 evaluation of ecosystem services. Functional diversity, a key aspect of biodiversity, provides a
23 more realistic characterisation of the functioning of ecological communities than only studying
24 their taxonomic diversity. The relevance of functional ecology studies has steadily increased
25 in agroecosystems. However, the combined effects of pesticides and land cover heterogeneity
26 on the taxonomic and functional diversity of arthropod communities have been studied less
27 frequently. We sampled arthropods during the dry season in 19 rice fields located in two
28 different regions of Northern Vietnam. We assorted the arthropods into functional groups
29 corresponding to different feeding habits and calculated the taxonomic and functional group
30 diversities. Finally, we analysed the impacts of pesticide applications and land cover
31 heterogeneity on both diversity measures. Taxonomic and functional group diversity measures
32 were highly correlated. In turn, both diversity measures responded similarly to land cover
33 heterogeneity and pesticides. Land cover heterogeneity had positive effects on taxonomic and
34 functional group diversity, mainly at the early stage of rice crops. Conversely, the impact of
35 pesticide application on both diversity measures was strongly negative. Our results suggest
36 that rice agroecosystems can be more sustainable by increasing landscape heterogeneity and
37 a reduced pesticide use. Such schemes may help to maintain higher levels of biodiversity that
38 ensure ecosystem functioning, which will be therefore likely beneficial to provide ecosystem
39 services in agroecosystems.

40 Keywords: Arthropod communities, diversity dimensions, land use intensity, natural enemies,
41 Northern Vietnam, rice agroecosystems

42

43 **1. Introduction**

44 Multiple hierarchical assembly processes have been shown to influence biodiversity patterns
45 across spatial scales. At broad spatial scales, for instance, dispersal among habitat patches is
46 key to the maintenance of diversity patterns in a metacommunity context (Hassan et al., 2016;
47 Oliver et al., 2010). At more local scales, abiotic filtering and biotic interactions play an
48 important role (De Bello et al., 2009; Gianuca et al., 2017; Kraft et al., 2015). Increasing human
49 impacts may alter such community assembly processes across scales and may lead to
50 significant biodiversity changes (Naeem et al., 2012). For instance, reduced connectivity in
51 fragmented landscapes influences dispersal rates among populations and communities.
52 Likewise, environmental degradation imposes a strong local filter on species traits and may
53 result in depauperated local communities (Brudvig et al., 2015; Haddad et al., 2015). Due to
54 well-established universal scale-dependence of biodiversity processes and patterns, it is ever
55 more recognised that community level analyses benefit from being conducted at multiple
56 spatial scales, especially if we want to correctly inform management and conservation
57 decisions (Hendrickx et al., 2007).

58 Diversity partitioning techniques have traditionally been applied to taxonomic-based
59 approaches, which are based on species identities but disregard their functional differences
60 (Hooper et al., 2002). It has been proposed that accounting for functional differences among
61 different ecological groups (Cardoso et al., 2014; Naeem et al., 2012) can provide a more
62 direct link between organisms and ecosystem processes (Cadotte, 2017; McGill et al., 2006).
63 In addition, phylogenetically distantly related organisms can have similar functionality due to
64 convergent evolution (Díaz et al., 2013). Therefore, focusing directly on functional diversity
65 instead of taxonomic or phylogenetic diversity can provide deeper insights into community
66 assembly (Cardoso et al., 2014; Hooper et al., 2002) and ecosystem functioning (De Bello et
67 al., 2010).

68 Land cover heterogeneity can support different ecological mechanisms, which have positive
69 effects on community structure (Oliver et al., 2010; Papanikolaou et al., 2017a, 2017b). For
70 example, higher landscape heterogeneity provides more opportunities for niche partitioning

71 (Amarasekare, 2003; Yang et al., 2015). Especially in agroecosystems, diverse land cover
72 types can act for different organisms as refuge (Gurr et al., 2017), food source (Westphal et
73 al., 2015) and overwintering site (Dufлот et al., 2015). Habitat heterogeneity may consequently
74 increase both taxonomic and functional diversity.

75 Land-use intensification is amongst the major drivers of biodiversity change (Foley et al., 2005;
76 Laliberté et al., 2010). It mostly results in a simplification of landscapes (e.g. large amount of
77 single, homogeneous types) and high external inputs (e.g. pesticides) (Foley et al., 2005;
78 Laliberté et al., 2010). This, in turn, leads to a loss of species in important groups, like
79 arthropods in agroecosystems (Hendrickx et al., 2007; Lingbeek et al., 2017), that provide
80 essential ecosystem services like biocontrol (Gurr et al., 2012, 2011). Yet, some studies have
81 demonstrated uncoupled responses of taxonomic and functional diversity to common
82 environmental drivers (e.g. De Palma et al., 2017), but it is not entirely clear how land-use
83 intensification, pesticide application, and habitat homogenization simultaneously influence
84 these two biodiversity measures (Mayfield et al., 2010; Peco et al., 2012). Some studies have
85 suggested that land-use intensification can lead to a reduction of functional group richness,
86 whereas taxonomic diversity can be more resilient against external disturbances (Schweiger
87 et al., 2007). This may happen when species with unique functional traits get replaced by
88 functionally redundant species (Ernst et al., 2006; Teresa and Casatti, 2012; Villéger et al.,
89 2010). Consequently, measuring functional diversity in addition to taxonomic diversity can help
90 to understand the full dimension of how anthropogenic land use and its change impacts
91 biodiversity (Ernst et al., 2006).

92 Here, we analyse how pesticides and land cover heterogeneity impacts both arthropod
93 taxonomic and functional group diversity. We focus on areas in Vietnam, where the
94 intensification of rice cultivation constantly increased during the last decades
95 (Schreinemachers et al., 2015). So far, the effects of pesticides and land cover heterogeneity
96 on taxonomic and functional group diversity of arthropods in rice agroecosystems in Vietnam
97 remain elusive and are poorly understood. Given the likely strong impacts of land-use
98 intensification on biodiversity and the potential uncoupled patterns of taxonomic and functional

99 group diversity, we hypothesise that increasing pesticide usage negatively influences
100 functional group diversity, although it may have a smaller effect on taxonomic diversity.
101 Furthermore, we hypothesise that higher land cover heterogeneity positively affects taxonomic
102 and functional group diversity.

103 **2. Material and Methods**

104 *2.1. Study area*

105 This study was part of the research project LEGATO (Land-use intensity and Ecological
106 enGineering-Assessment Tools for risks and Opportunities in irrigated rice based production
107 systems; Settele et al., 2018) and was carried out in 19 rice fields located in two rice dominated
108 lowland regions along the Red River Delta in Northern Vietnam. In these areas, rice is the main
109 crop (Global Rice Science Partnership, 2013). The Red River Delta is characterised by a warm,
110 humid, and subtropical climate (Klotzbücher et al., 2015) with a distinct seasonality with two
111 growing seasons per year. The first season ranges from February to May and the second from
112 July to October (Klotzbücher et al., 2015).

113 The first region, Hai Duong (LEGATO region VN1: 21°00'N 106°23'E), is situated 60 km east
114 of Hanoi. The region is heavily industrialised and dominated by intensively farmed rice fields.
115 We originally selected ten rice fields but in one of these fields the cropping system changed
116 during the investigation period and we removed it from the analysis.

117 The second region, Vinh Phuc (LEGATO region VN2: 21°20'N 105°43'E), is located 35 km
118 northwest of Hanoi. Similarly to Hai Duong, the landscape is dominated by rice fields but is
119 industrialised to a lesser extent (Burkhard et al., 2015). In this region, we selected ten rice
120 fields.

121 The average distance between rice fields was 338 m within a region. The size of the rice fields
122 was on average 491 m² ranging from 97 - 1883 m² (Appendix: Table A.1). All investigated rice
123 fields were sprayed with pesticides and fertilised using chemically produced NPK (Nitrogen,
124 Phosphorus, Potassium) fertiliser during our study (Klotzbücher et al., 2015). In Hai Duong,
125 farmers use rice varieties that are highly productive, whereas in Vinh Phuc farmers use

126 traditional varieties with higher genetic diversity (Burkhard et al., 2015). Similar to Wilby et al.
127 (2006) and Dominik et al. (2017), all observations and investigations were implemented in real
128 agricultural settings without controlling external factors. The decisions about agricultural
129 practices, like fertiliser use, weeding, pesticide application, and the choice of rice varieties were
130 left to the farmers.

131 *2.2. Arthropod sampling and assignment*

132 We collected arthropods during the dry season from March to April in 2015 using blow vac and
133 sweep net. Both are highly effective standard methods to sample arthropods or specific taxa
134 in rice agroecosystems (Bambaradeniya et al., 2004; Gangurde, 2007; Ghahari et al., 2008;
135 Schoenly et al., 2010). We sampled arthropods during the vegetative stage of the rice plant at
136 two points in time: 35 and 50 days after the rice seedlings were transplanted into the fields
137 (days after transplanting = DAT). Sampling times are in accordance with the overall LEGATO
138 sampling design in order to standardise methodologies. For each sampling time, we took five
139 replicates per method in each rice field. For sweep net, we sampled arthropods at two locations
140 along the rice bunds and at three locations within the centre of each rice field. For each sample
141 location, we sampled an area of 30 m² while walking with a speed of approximately 0.5 m/sec
142 and performing 30 sweeps. If the size of a rice field was too small to locate five sample units,
143 we took the remaining units in the immediate vicinity (i.e. in the neighbouring field). Sample
144 units did not overlap with one another. With the blow vac method, we sampled an area of 0.25
145 m² in five randomly chosen locations within a rice field using a square plastic enclosure fitted
146 with nylon net on top to prevent arthropods from escaping.

147 Arthropods were counted, identified to family level, and assigned to functional groups, which
148 are based on similar functional behaviour and food acquisition strategies. We used the
149 following functional groups defined after Shepard et al. (1995, 1987) and Heong et al. (1991):
150 predators, parasitoids, herbivores, decomposers (detritivores and scavengers), and
151 fungivores. Arthropod samples that could not be assigned to one of these functional groups,

152 due to the samples' poor condition, or could be assigned to more than one group (a total of
153 0.001% of all collected arthropods) were excluded from further analysis.

154 *2.3. Predictor variables*

155 *2.3.1. Land cover heterogeneity*

156 To investigate the relationship between land cover heterogeneity and arthropod taxonomic and
157 functional group diversity, we recorded and classified the land cover within a 300 m buffer
158 around the centre of the rice field. Land cover was classified based on digital habitat mapping
159 by satellite images following Burkhard et al. (2015). Land cover was classified into the following
160 ten types: bare soil, forest, fruit, meadow/grassland, rice field, vegetable, water, crops,
161 compacted surface and sealed surface (see Appendix: Table A.1 and A.2 for more
162 information). We calculated the Shannon diversity (H') index to measure land cover
163 heterogeneity based on the proportion of land cover types (in %) within the 300 m buffer.

164

165 *2.3.2. Pesticides*

166 To investigate the relationship between pesticides and arthropod functional and taxonomic
167 groups, we assessed the number of all pesticide applications combined and of insecticide
168 applications separately by interviewing farmers. Farmers used herbicides, fungicides,
169 molluscicides, rodenticides and insecticides which are here summarised as pesticides.
170 Detailed information about the used pesticides can be found in Sattler et al. (2018). Farmers
171 sprayed pesticides on average four times (ranging from four to six) during the cropping season
172 and all farmers sprayed pesticides immediately after transplanting the rice plants (Sattler et al.,
173 2018).

174

175 *2.4. Data analysis*

176 For statistical analysis, we used abundance data of either the functional groups or the
177 taxonomic units of arthropod communities of each sample unit per rice field from both sweep
178 net and blow vac sampling (five replicates per method, field and at 35 and 50 DAT). Diversity

179 indices were calculated using Shannon entropy (H' , Shannon-Wiener index). Due to the
180 difficult interpretation of most standard indices (Jost, 2007; Marcon and Hérault, 2015), we
181 converted the Shannon entropy to the 'effective number of species' by taking its exponential
182 (Jost, 2007, 2006), which is referred to as 'effective number of functional or taxonomic groups'.
183 For simplification, the term 'diversity' will be used instead of 'effective number of functional
184 groups or taxonomic groups'.

185 We calculated the local (= alpha) diversity of each rice field (Marcon and Hérault, 2015). Alpha
186 diversity was defined as the diversity of functional or taxonomic groups of all specimens found
187 in one rice field. We also partitioned diversity of arthropod communities at the landscape scale,
188 whereby gamma was defined as the diversity of functional or taxonomic groups of all
189 specimens found in one region, alpha as the local communities within a rice field (i.e.
190 aggregated across the sampling units) of one region and beta as the variation of taxonomic
191 and functional compositions among the rice fields within a region. Scale definitions (local and
192 landscape scale) are according to Willis and Whittaker (2002).

193 Data of taxonomic and functional group diversity were normally distributed (tested with the
194 Shapiro-Wilk test; Royston 1982). To analyse the relationship of pesticide applications (for
195 both all pesticides and insecticides only) and land cover heterogeneity with the calculated
196 taxonomic and functional group diversity we used linear mixed effects models. To control for
197 potential non-independence of the data points, we included rice fields nested within region as
198 random effects. The number of pesticide applications and land cover heterogeneity were not
199 collinear (Pearson correlation $r=0.12$) and thus included as fixed effects in the models. For
200 these analyses, only the alpha diversity at the local scale was used because of too few data
201 points at the landscape scale for beta and gamma diversity (only one data point) per region.

202 Model selection followed a multimodel inference approach relying on the second-order Akaike
203 information criterion (AICc; Burnham and Anderson 2002) and only the best model was
204 selected. Candidate models included as predictor variables: number of pesticide applications
205 and land cover heterogeneity. The variable importance was computed based on the sum of
206 AICc weights for each model in which the predictor variable appeared.

207 In total, we performed eight linear mixed effects models separated by methods (sweep net and
208 blow vac), sampling days (35 and 50 DAT), and for both functional and taxonomic measures.
209 We calculated for each model the marginal and conditional R-squared values. To compare the
210 effect size (calculated marginal R^2) of pesticides on taxonomic and functional group diversity,
211 we used the Paired Student's t test as data were normally distributed (tested with the Shapiro-
212 Wilk test). Pairs were separated into methods and sampling days. Paired Student's t test was
213 also used to compare the taxonomic and functional group diversity between the two sampling
214 days (35 and 50 DAT). All analyses were performed in the statistical environment R for
215 Windows (Version 3.2.4; R Core Team 2016) using the packages 'lmer4' (Bates et al., 2015),
216 'MuMIn' (Bartoń, 2016), 'vegan' (Oksanen et al., 2017), 'ggplot2' (Wickham, 2015), 'entropart'
217 (Marcon and Héroult, 2015), and 'lmerTest' (Kuznetsova et al., 2017).

218

219 **3. Results**

220 Overall, we collected 164,671 arthropod specimens belonging to 17 orders and 77 identified
221 families. In all functional groups the highest number of specimens was sampled by sweep
222 netting, with 151,585 specimens. Blow vac sampling yielded a total of 13,086 specimens. We
223 collected 73 different families with the sweep netting method and 57 families with the blow vac
224 method. A comparison of collected functional groups by the two sampling methods can be
225 found in Fig. 1. Arthropods were dominated by the class Insecta with 99.8%. The class of
226 Arachnida was present with 0.2% and one specimen of Diplopoda was found. Decomposers
227 were the most abundant functional group with 132,662 specimens followed by herbivores with
228 19,697 specimens, predators with 7,321 specimens, parasitoids with 4,720 specimens, and
229 fungivores with 58 specimens. In total, 213 specimens could not be assigned into a specific
230 functional group and were excluded from further analysis. A list of all collected arthropods can
231 be found in the Appendix (Table A.3).

232 Generally, taxonomic and functional group diversity were highly correlated (Pearson
233 correlation $r = 0.94$). On the landscape scale, we found very low beta diversity among the rice
234 fields for both taxonomic and functional group diversity (see Table 1 for landscape scale).

235 We found higher functional group diversity at 50 DAT compared to 35 DAT for both sampling
236 methods (t-test; Blow vac: $t = -3.87$, $df = 18$, $p\text{-value} = 0.001$; sweep net: $t = -4.04$, $df = 17$, $p\text{-}$
237 $value = 0.0008$). A similar result was found for taxonomic diversity (t-test; Blow vac: $t = -2.87$,
238 $df = 18$, $p\text{-value} = 0.01$; sweep net: $t = -2.52$, $df = 18$, $p\text{-value} = 0.02$).

239

240 *3.1. Relationship of land cover heterogeneity and pesticides with functional and taxonomic* 241 *group diversity at local scale*

242 Given the high correlation between functional and taxonomic diversity, the impacts of
243 pesticides and land cover heterogeneity were highly similar for both diversity measures. Alpha
244 diversity decreased with an increasing number of pesticide applications consistently across
245 sampling methods and dates (Fig. 2a, b, c, e; Fig. 3a, b, c, e). However, an effect of land cover
246 heterogeneity on alpha diversity was only evident for communities sampled with sweep nets
247 at 35 DAT where diversity increased with increasing land cover heterogeneity (Fig. 2d; Fig 3d).
248 Linear mixed effects models at local scale can be found in Table 2 and Table 3. The global
249 models which include all predictor variables can be found in the Appendix (Table A.4, Table
250 A.5 and Figure A.1).

251 The effect size (marginal R^2) of pesticides on taxonomic and functional group diversity showed
252 no differences based on the Paired Student's t test ($p > 0.05$). Therefore, the relationship of
253 pesticides with both diversity dimensions (taxonomic and functional group diversity) was
254 negative to a similar extent. The number of insecticide applications alone did not result in
255 significant effects for both taxonomic and functional group diversity. Results of global models
256 of insecticide applications and land cover heterogeneity can be found in the Appendix (Table
257 A.6 and A.7).

258

259

260 **Table 1 Taxonomic and functional group diversity (mean of alpha, beta, and gamma diversity) at**
 261 **landscape scale.** Functional and taxonomic alpha (α), beta (β), and gamma (γ) diversity for Hai Duong
 262 (VN1) and Vinh Phuc (VN2), separated for blow vac (BV) and sweep net (SN) as well as for sampling
 263 days (35 and 50 days after transplanting).

Method	Reg	α (functional)	β (functional)	γ (functional)	α (taxonomic)	β (taxonomic)	γ (taxonomic)
BV 35	VN1	2.16	1.02	2.20	4.14	1.35	5.57
	VN2	2.04	1.05	2.15	3.46	1.32	4.58
BV 50	VN1	2.82	1.08	3.06	5.65	1.41	7.98
	VN2	2.48	1.18	2.94	4.93	1.54	7.61
SN 35	VN1	1.83	1.02	1.87	2.47	1.07	2.65
	VN2	1.72	1.05	1.81	2.24	1.09	2.45
SN 50	VN1	3.18	1.05	3.32	5.53	1.18	6.52
	VN2	2.08	1.03	2.13	2.86	1.10	3.15

264

265 **Table 2 Linear mixed effects models of functional group diversity at local scale (alpha diversity).**

266 The best model was selected among the candidate models following a multimodel inference approach
 267 separately for blow vac (BV) and sweep net (SN) as well as sampling days (35 and 50 days after
 268 transplanting). R-squares are shown as marginal R-squared values (R^2m) and conditional R-squared
 269 values (R^2c).

Model	Response variable	Predictor variable	p-value	Variable importance	R^2m	R^2c
BV 35	alpha diversity	Number of pesticide applications	0.005	0.59	0.31	0.52
BV 50	alpha diversity	Number of pesticide applications	0.02	0.43	0.26	0.26
SN 35	alpha diversity	Number of pesticide applications	0.003	1	0.46	0.46
		Land cover heterogeneity	0.05	1		
SN 50	alpha diversity	Number of pesticide applications	0.05	0.29	0.18	0.55

270

271 **Table 3 Linear mixed effects models of taxonomic diversity at local scale (alpha diversity).** The

272 best model was selected among the candidate models following a multimodel inference approach
 273 separately for blow vac (BV) and sweep net (SN) as well as sampling days (35 and 50 days after
 274 transplanting). R-squares are shown as marginal R-squared values (R^2m) and conditional R-squared
 275 values (R^2c).

Model	Response variable	Predictor variable	p-value	Variable importance	R^2m	R^2c
BV 35	alpha diversity	Number of pesticide applications	0.01	0.69	0.27	0.49
BV 50	alpha diversity	Number of pesticide applications	0.03	0.71	0.25	0.25
SN 35	alpha diversity	Number of pesticide applications	0.008	0.68	0.37	0.5
		Land cover heterogeneity	0.09	0.3		
SN 50	alpha diversity	Number of pesticide applications	0.05	0.3	0.05	0.89

276

277 **4. Discussion**

278 *4.1. Effects of pesticides and land cover heterogeneity on taxonomic and functional group*
279 *diversity*

280 Beta diversity at the regional scale was consistently low for both sampling methods and
281 sampling days. This indicates that arthropod communities do not differ much among the
282 different rice fields of a region and that local arthropod diversity is largely defined by the
283 regional species pool. However, the observed level of variation among the rice fields in a region
284 was strongly related to the number of pesticide applications indicating that both taxonomic and
285 functional group diversity of arthropod communities decrease with pesticide applications in the
286 rice fields. Insecticide applications alone did not affect taxonomic and functional group
287 diversity, which shows that the combination of all applied pesticides have stronger effects on
288 the taxonomic and functional group diversity than insecticide applications only. This was
289 similarly reviewed by Wu et al. (2020), who pointed out that insecticides and the combination
290 of pesticides can lead to two different effects: acute vs. chronic pest resurgence. The acute
291 resurgence is caused by a higher sensitivity of natural enemies to insecticides compared to
292 pest species. Consequently, the higher mortality of natural enemies stimulates pest
293 reproduction. Chronic resurgence of pests, on the other hand, emerges if a combination of
294 pesticides has smaller effects on natural enemies but positively induces pest reproduction at
295 longer latency. Since we did not find evidence for an insecticide-only effect but arthropod
296 diversity rather reacted to all pesticides in combination, arthropod communities are likely more
297 affected by chronic pest resurgence in our study region.

298 Heterogeneous land cover in the areas surrounding rice fields was associated with an increase
299 in functional and taxonomic arthropod diversity. However, this effect was only found in the early
300 rice growth stage and only when using the sweep net sampling method.

301 Taxonomic and functional diversity can be closely connected but do not necessarily need to
302 correlate (Cardoso et al., 2014; Mayfield et al., 2010). Flynn et al. (2009) studied the effect of
303 land-use intensity on mammals, birds, and plants. Similar to our study, both species richness

304 and functional diversity declined with land-use intensity. However, Peco et al. (2012) studied
305 the effect of grazing abandonment on functional and taxonomic diversity of grasslands and
306 found a loss of functional diversity rather than species richness. In a study by Schweiger et al.
307 (2007), increasing land-use intensity led to decreasing functional richness of hoverfly
308 communities rather than affecting species richness. Villéger et al. (2010) showed contrasting
309 responses of biodiversity in aquatic ecosystems influenced by habitat degradation: functional
310 diversity of fish was negatively affected whereas fish species richness increased. In our study,
311 taxonomic and functional group diversity were highly correlated, which is reflected by similar
312 responses to pesticides and land cover heterogeneity. Taxonomic and functional group
313 diversity consistently showed a negative response to the number of applied pesticides
314 regardless of the sampling method. Our hypothesis that increasing pesticide use negatively
315 affects functional group diversity can be confirmed. However, we cannot confirm our
316 hypothesis that increasing pesticide usage would have a smaller effect on taxonomic diversity
317 than on functional group diversity, since the effects were similarly strong for both taxonomic
318 and functional group diversity. This might be either because each species is broadly
319 functionally unique or because of the non-independence in our two diversity metrics. Despite
320 of the potential non-independence of the two diversity metrics, we expected different results if
321 a single or few functional groups had been lost but compensated by an increasing number of
322 species from different taxonomic groups, therefore reducing the impact on taxonomic diversity
323 while still affecting functional group diversity. This mechanism was shown by Ernst et al.
324 (2006), who found a negative effect of forest degradation on functional diversity of amphibians
325 but no effect on taxonomic diversity.

326 Furthermore, we hypothesised that land cover heterogeneity would increase taxonomic and
327 functional group diversity. In general, we only found effects of land cover heterogeneity with
328 sweep net data at 35 DAT. A study by Wilby et al. (2006), which focused on similar questions,
329 found under 'real' agricultural conditions that arthropod species diversity in rice fields generally
330 decreases with a decrease in structural diversity in the surroundings, which is in line with our
331 results for 35 DAT using the sweep net sampling method. One reason that we did not find a

332 similar effect for blow vac data could be that the probabilities of sampling particular species
333 groups likely differ between both methods. Indeed, sweep net data contained a higher
334 abundance of different taxonomic and functional groups compared to blow vac samples. This
335 might increase the likelihood to cover communities that are more responsive to habitat
336 heterogeneity.

337 Taxonomic and functional group diversity and abundance can change with crop age (Wilby et
338 al., 2006) and these different communities might respond differently to land cover
339 heterogeneity depending on whether they migrate from somewhere else in the landscape at a
340 given point in time. For instance, early arriving arthropod groups mainly immigrate into the rice
341 fields from the surroundings (Settle et al., 1996; Wilby et al., 2006), and thus fine-scale land
342 cover heterogeneity seems to benefit these groups, as our results have shown. Dominik et al.
343 (2018) showed that the effects on a very local scale can be important: bunds (levee of
344 terrestrial area surrounding the fields) build an extensive network connecting the rice fields.
345 Such bunds often have sparse vegetation that can potentially offer alternative food resources
346 or refuge to natural enemies (Way and Heong, 1994) and likely facilitate the ability of rice
347 arthropods to move through the rice agroecosystem (Sigsgaard, 2000; Yu et al., 1996; see
348 Settele and Settle, 2018 for further discussion).

349 Nevertheless, such an effect might have diminished at a later stage of rice plants (50 DAT)
350 when arthropod composition changes. Dominik et al. (2017) found no effect of fine-scale
351 landscape heterogeneity on assemblage structure of arthropod communities in rice fields in
352 the Philippines. They argue that regional-scale effects like climate conditions, elevation and
353 landscape structure at broader scale might be more important than fine-scale effects. Another
354 reason for the diminishing effect of land cover heterogeneity on arthropod diversity at the later
355 stage might be an increasing pesticide application in the surroundings. Many of the rice fields
356 in our study were surrounded by fruit and vegetable fields which can suffer from even higher
357 pesticide applications compared to rice fields (Hoi et al., 2016; Van Mele et al., 2002). This
358 means that not only pesticide application within rice fields influences the arthropod
359 communities in the rice fields, but also the application of pesticides in the surrounding non-rice

360 habitats may counteract the positive effects of land cover heterogeneity. Thus, management
361 practices in the surrounding land-use types might be important drivers of the diversity in rice
362 agroecosystems; an important research question for future studies.

363

364 *4.2. Outlook and Conclusion*

365 Rice ecosystems depend on multiple functions related to multiple ecosystem services, such
366 as pest control by predators and parasitoids or nutrient cycling mediated by decomposers
367 (Schmidt et al., 2016). A decline of functional diversity can lead to a loss of ecosystem services
368 (Villéger et al., 2010). Also, a change in taxonomic composition can impact ecosystem
369 processes as even single species can hold key functions necessary for a stable ecosystem
370 (Chapin et al., 2000; Hooper et al., 2002). Therefore, the maintenance of taxonomic and
371 functional group diversity is important for rice agroecosystems. However, our study showed
372 that high levels of pesticide applications lead to a reduction of the two diversity dimensions,
373 while land cover heterogeneity can have a positive effect. Gurr et al. (2011) showed that land
374 cover heterogeneity in rice-based landscapes can be improved by ecological engineering. For
375 instance, when rice bunds are planted with flowering plants they can provide additional food
376 sources, such as nectar and pollen, as well as shelter for arthropods from various functional
377 groups (Hassan et al., 2016). Further studies might focus on whether land cover heterogeneity
378 and pesticides have an effect on single functional groups and if there is a shift in rice yield
379 when reducing pesticides and increasing landscape heterogeneity. Previous studies
380 successfully introduced ecological engineering to farmers in the Mekong Delta (e.g. Heong et
381 al. 2014; Le 2014), but field studies are mostly local and there is no law enforcement to
382 implement ecological engineering by farmers. To obtain more sustainable farming practises in
383 rice agroecosystems, approaches like ecological engineering should be more in the focus of
384 research studies.

385

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393

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645 **Fig. 1. Total abundance of functional groups.** Total abundance (log-transformed) of decomposers,
646 herbivores, parasitoids, and predators sampled by using blow vac (BV) and sweep net (SN) at 35 a) and
647 50 b) days after transplanting (DAT).

648

649 **Fig. 2. Functional group diversity: Relationship between the number of pesticide applications (a-**
650 **c, e) and land cover heterogeneity (d) for blow vac (BV) and for sweep net data (SN) at 35 and 50**
651 **days after transplanting.** Alpha diversity is based on the exponential Shannon entropy and expressed
652 as effective numbers of functional groups. Pesticides were measured as number of pesticide
653 applications. Land cover heterogeneity is based on the proportion of habitat types and was calculated
654 with the Shannon index (H').

655

656 **Fig. 3. Taxonomic diversity: Relationship between the number of pesticide applications (a-c, e)**
657 **and land cover heterogeneity (d) for blow vac (BV) and for sweep net data (SN) at 35 and 50 days**
658 **after transplanting.** Alpha diversity is based on the exponential Shannon entropy and expressed as
659 effective numbers of taxonomic groups. Pesticides were measured as number of pesticide applications.
660 Land cover heterogeneity is based on the proportion of habitat types and was calculated with the
661 Shannon index (H').

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