

1 **Title page**

2 Trophic niche structure of invertebrate-eating bats and birds in West African rice agroecosystems
3 revealed by DNA metabarcoding

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22 **Abstract**

23 Knowledge of animal community diets is essential for understanding ecosystem functioning. Bats and
24 birds are important groups of invertebrate predators, managing their populations. However, the diets of
25 West African species and the mechanisms shaping them remain poorly understood. In this study, we
26 investigated these mechanisms in agricultural landscapes of Guinea-Bissau using metabarcoding and
27 next-generation sequencing. Specifically, we asked: 1) How do dietary breadth and composition vary
28 among predator species and between bats and birds 2) How does dietary overlap vary within and
29 between guilds? (3) To what extent is dietary dissimilarity associated with phylogeny?

30 We analysed the diet of 13 bat and eight bird species. Results revealed that Hemiptera (18.5%);
31 Coleoptera (16.6%), Blattodea (15.8%), Lepidoptera (15.5%) and Orthoptera (11.4%) were the most
32 consumed orders, with termites (Blattodea, Termitidae) standing out (8.2%). Based on OTU richness
33 and sample-based rarefaction, all predators exhibited broad niches (99.7 ± 37.4 OTUs per predator
34 species, on average), with bats generally exhibiting broader niches. Bats' and birds' diets differed
35 significantly (ANOSIM $R = 0.61$, $p = 0.001$), possibly driven by differences in foraging periods. At the
36 OTU level, birds displayed significant dietary segregation ($O_{\text{birds}} = 0.05$, $p = 0.017$), whereas some bats
37 displayed significant dietary overlap ($O_{\text{bats}} = 0.23$, $p = 0.001$), indicating low interspecific competition.
38 However, we found no strong correlation between diet dissimilarity and phylogeny.

39 These findings provide insight into niche partitioning and may inform future assessments on the
40 importance of the ecological role of bats and birds in West African agroecosystems.

41 **Key words**

42 Dietary overlap; Community ecology; Niche partitioning; Ecosystem services; Biodiversity
43 conservation

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47 **Introduction**

48 Diet is a fundamental component of animal ecology because it defines the resources that organisms use
49 and therefore constrains the ecological niche they can occupy (McCann 2007). Dietary studies provide
50 insights into species' energetic requirements, ecological roles, and the trophic interactions that structure
51 communities. Additionally, studies of trophic niches provide essential insights into resource partitioning
52 and coexistence, particularly in species-rich assemblages where multiple consumers exploit overlapping
53 prey pools (Schoener 1974; Novella-Fernandez et al. 2020). However, trophic niches are rarely
54 structured along a single ecological axis; instead, multiple mechanisms may operate simultaneously and
55 at different hierarchical levels within predator assemblages (Chesson 2000; McGill et al. 2006)

56 Bats and birds are two major groups of aerial insectivorous that exert strong top-down effects on
57 invertebrate communities, thereby influencing ecosystem functioning across terrestrial ecosystems
58 (Kunz et al. 2011; Nyffeler et al. 2018). Their trophic ecology has been shown to vary widely among
59 species, shaped by morphological, behavioural, and environmental differences (Salinas-Ramos et al.
60 2015; Maine and Boyles 2015; Mansor et al. 2018; Davies et al. 2022). Although regional context
61 modifies prey availability and predator foraging opportunities, a consistent global pattern is that
62 insectivorous bats generally display broader diets than most birds (Maas et al. 2016). This difference is
63 frequently attributed to their nocturnal activity, the greater taxonomic diversity of nocturnal arthropods,
64 and the broad range of foraging strategies adopted by bats (Maas et al. 2016; Wong and Didham 2024).
65 Temporal segregation of foraging activity therefore represents one important axis of niche differentiation
66 among aerial predators (Kronfeld-Schor and Dayan 2003), but its relative contribution remains unclear
67 when considered alongside other ecological and evolutionary factors. Evolutionary history may also
68 contribute to trophic divergence, with more distantly related species often differing in diet due to
69 morphological constraints, sensory adaptations, and distinct ancestral foraging strategies (Van
70 Cakenberghe et al. 2002; Brändle et al. 2002; Carrillo-Araujo et al. 2015). Conversely, closely related
71 species may display similar diets unless competition or behavioural differentiation promotes trophic
72 segregation (Schoener 1974; Stevens and Willig 2000), as predicted by traditional niche theory
73 (Macarthur and Levins 1967). Nevertheless, dietary similarity does not necessarily scale linearly with

74 phylogenetic distance, particularly in highly mobile and opportunistic predators, where behavioural
75 plasticity and fluctuating resource availability may override evolutionary proximity.

76 Tropical systems, characterised by high arthropod diversity and pronounced seasonal variability, provide
77 a valuable context in which to examine how multiple ecological axes structure trophic niches (Novotny
78 and Miller 2014). Agricultural mosaics such as rainfed rice systems combine high productivity with
79 marked temporal changes in prey availability (Lawler 2001), while still maintaining diverse predator
80 communities (Chaves et al. 2026). These landscapes therefore offer an opportunity to examine how
81 temporal activity, trophic strategy, and evolutionary history interact in shaping dietary patterns within
82 and between predator guilds.

83 Thus, trophic differentiation in aerial predator assemblages may emerge from distinct ecological axes
84 operating at different hierarchical levels. Temporal activity may structure broad guild-level segregation,
85 whereas within-guild dietary patterns may be shaped by behavioural and ecological traits rather than by
86 phylogenetic relatedness alone. At the same time, substantial dietary overlap may arise where multiple
87 predators exploit a broadly shared prey pool, making it important to distinguish between differences in
88 prey composition and complete trophic segregation.

89 In this study, we used DNA metabarcoding and next-generation sequencing (NGS) to examine trophic
90 niche structure in a community of invertebrate-eating bats and birds foraging in West African rice
91 landscapes. Specifically, we addressed the following questions: 1) How do dietary breadth and
92 composition vary among predator species and between major predator guilds (here defined as nocturnal
93 insectivorous bats and predominantly diurnal birds)? 2) How does dietary overlap vary within and
94 between guilds? (3) To what extent is dietary dissimilarity associated with phylogenetic distance across
95 the assemblage?

96 We hypothesized that trophic niches in this assemblage are structured hierarchically along multiple
97 ecological axes. Specifically, we predict that (i) all predator species exhibit relatively broad arthropod
98 diets consistent with the high prey diversity characteristic of tropical ecosystems (Novotny and Miller
99 2014); (ii) temporal activity patterns contribute to trophic differentiation between major predator guilds,

100 leading to differences in dietary composition between nocturnal bats and diurnal birds; (iii) phylogenetic
101 relatedness does not consistently predict dietary similarity, either within guilds or across the assemblage,
102 suggesting that ecological and behavioural traits mediate trophic overlap at finer scales.

103 **Materials and methods**

104 **Study area**

105 The study was conducted in the Oio region of northern Guinea-Bissau (Figure 1), an area characterised
106 by a Guinean maritime climate with two distinct seasons, a rainy season from May to November, and a
107 dry season from December to April (FAO and ICRISAT 2019). Mean annual air temperature is about
108 28°C with little seasonal variation and annual precipitation ranges from 1200 to 1400 mm (Catarino et
109 al. 2001; Climate Change Knowledge Portal 2023). The dominant vegetation is savannah woodland, a
110 secondary formation largely shaped by human interventions such as fires and shifting agriculture
111 (Catarino et al. 2008). The wider landscape also includes extensive cashew orchards, small villages
112 (*tabancas*) and mosaics of smallholder agricultural plots.

113 Our work focused on freshwater lowland rainfed rice fields within the Mansabá sector (12° 20' 0.314"
114 N, 15° 10' 57.474" W; Figure 1), locally known as *bolanhas*. These fields occupy former wet grass
115 savannahs that flood during the rainy season (Catarino et al. 2008), creating seasonally inundated
116 habitats that support high invertebrate and vertebrate diversity (Lawler 2001). These rice systems
117 comprise actively cultivated paddies interspersed with fallow areas and isolated trees and are bordered
118 by savannah woodlands and cashew orchards. Invertebrate communities in these landscapes include
119 numerous species that are often considered important rice pests (Heinrichs and Barrion 2004), some of
120 which can cause substantial crop damage (Oerke 2006; Waddington et al. 2010).

121 In Guinea Bissau, rice cultivation is a central component of rural livelihoods, and rice fields cover
122 approximately 21% of the country's harvested area (FAO and ICRISAT 2019). Reflecting its importance,
123 rice production has expanded over recent decades (Djata et al. 2003; USDA 2024), a trend mirrored
124 across several African countries (Seck et al. 2012).

125 **Bat and bird sampling**

126 Fieldwork was conducted from June to December in 2021 and 2022, spanning the entire rice growth
127 cycle, except for August when sampling was suspended due to heavy rainfall. Captures were performed
128 using 12 m mist nets (Ecotone, 712/2P). On average, five nets were deployed per sample session (range:
129 1-8), positioned along rice field margins, near isolated trees, and close to water bodies to maximize
130 capture success for both bats and birds. Nets were operated from 07:00 to 11:00 and from 17:00 to 22:00,
131 covering the main activity periods of both groups.

132 Captured animals were kept in single-use paper bags for at least 30 minutes to allow for faecal sample
133 collection. Faecal pellets were collected from the holding bags, with up to three pellets obtained per
134 individual. Each pellet was placed in a separate 2 ml tube containing 96% ethanol. Captured individuals
135 were then identified to the lowest taxonomic level possible, and standard morphometric measurements
136 were taken. Bird measurements included wing, weight, and tarsus, while bat measurements included
137 forearm length, weight, and sexual characteristics. When possible, sex and age were also determined.
138 Bird identification followed Borrow and Demey (2014), with nomenclature following Clements (2007).
139 Bat identification followed Happold and Happold (2013), with nomenclature following Simmons and
140 Cirranello (2025). When field identification was uncertain, it was confirmed through molecular
141 techniques (see Section “Laboratory procedures”). All captures and handling procedures were conducted
142 under permits issued by the Instituto da Biodiversidade e Áreas Protegidas (IBAP), the national
143 regulatory authority in Guinea-Bissau, and complied with applicable ethical guidelines for the use of
144 wild animals in research.

145 **Laboratory procedures**

146 We individually processed and extracted the DNA of 665 faecal pellets, of which 526 belonged to bats
147 and 139 to birds. The number of individuals and pellets analysed was determined by balancing the
148 maximization diet estimation accuracy with the associated costs (Mata et al. 2019). DNA extraction
149 from bat faecal samples followed the protocol described by (Gonçalves et al. 2024). Initially, 650 µL of
150 lysis buffer (0.1 M Tris-HCl, 0.1 m EDTA, 0.01 M NaCl, 1% N- lauroylsarcosine, pH 7.5–8; Maudet et
151 al. 2002) was added to each sample, followed by homogenization with a sterile spatula, vortexing, and
152 incubation in a dry bath for 30 min at 56°C. Samples were vortexed again for 1 minute and centrifuged

153 at 12,000 × g for 30 seconds. Up to 500 µL of supernatant was then transferred to a new tube, and 25 uL
154 of OB Protease was added. The remaining steps followed the E.Z.N.A. Tissue Kit (Omega Bio-Tek,
155 Georgia, USA) protocol, except that DNA was eluted twice in 50 µL to create two extracts. For bird
156 faecal samples, we followed the manufacturer’s protocol from the Stool DNA Isolation Kit (Norgen
157 Biotek Corporation, Ontario, Canada). Extractions were conducted in batches of 23 samples, each
158 including one negative control without faecal sample. The extracted DNA was placed in 96-well plates
159 along with the negative controls, following the order of extraction. The last well in each plate was left
160 empty for negative control of the polymerase chain reaction (PCR).

161 The prey DNA was amplified using the FwhF2-R2n COI primers (Vamos et al. 2017), modified to
162 include Illumina adaptors (FwhF2: 5’ -
163 TCGTCGGCAGCGTCAGATGTGTATAAGAGACAGGGDACWGGWTGAACWGTWTAYCCHCC
164 - 3’; FwhR2n: 5’ -
165 GTCTCGTGGGCTCGGAGATGTGTATAAGAGACAGGTRATWGCHCCDGCTARWACWGG -
166 3’). These primers have proven effective in amplifying a broad range of arthropods and perform well
167 with degraded DNA (Elbrecht et al. 2019; Mata et al. 2021; da Silva et al. 2024). The PCR reaction
168 mixture consisted of 5 µL of Qiagen Multiplex Master Mix, 0.3 µL of each primer at 10 nmol/L, 3.4 µL
169 of water, and 1 µL of extracted DNA. The cycling conditions included 15 minutes at 95°C, followed by
170 40 cycles of 30 seconds denaturation at 95°C, 30 seconds of annealing at 50°C, and 30 seconds
171 elongation at 72°C, concluding with a final elongation step of 10 minutes at 72°C.

172 The field IDs of bats and birds were verified by amplifying a small fragment of the COI gene using the
173 FwhF1-R1 primers (Vamos et al. 2017), also modified with Illumina adaptors (FwhF1: 5’ -
174 TCGTCGGCAGCGTCAGATGTGTATAAGAGACAGYTCHACWAAYCAYAARGAYATYGG - 3’;
175 FwhR1: 5’ -
176 GTCTCGTGGGCTCGGAGATGTGTATAAGAGACAGARTCARTTWCCRAAHCCHCC - 3’). This
177 step confirmed the identity of cryptic species, as the Fwh2 primers often fail to detect vertebrate taxa.
178 The PCR conditions for this amplification were the same as those used for the prey DNA. After
179 amplification, the PCR products were diluted 1:4. A second PCR was performed to add 7bp long indexes

180 along with the Illumina P5 and P7 sequencing adaptors. The PCR mix and cycling parameters were
181 similar to the first PCR, but KAPA HiFi HotStart ReadyMix (Rocher, Basel, Switzerland) was used with
182 eight cycles of denaturation and annealing at 55°C, and extension. The indexed PCR products were
183 purified using Agencourt AMPure XP beads (Beckman Coulter, Brea, California, USA) at a 1:0.8 ratio.
184 The purified products were quantified using an Epoch Microplate Spectrophotometer (Agilent,
185 California, USA), diluted to a concentration of 15 nM, and pooled per maker. Each library was
186 individually quantified using qPCR (KAPA Library Quant Kit qPCR Mix; Rocher), diluted to 4 nM, and
187 sequenced on a MiSeq targeting a depth of 30,000 reads for Fwh2 and 4000 reads for Fwh1.

188 **Bioinformatic analysis**

189 The software Flash was used to merge paired reads (Magoč and Salzberg 2011), considering a minimum
190 expected overlap of 10 and maximum of 250 bp. Obitools commands 'ngsfilter,' 'obiuniq,' and 'obigrep'
191 (Boyer et al. 2016) were employed to, respectively, remove primer sequences, dereplicate reads within
192 each sample, and discard singletons per sample. VSEARCH was then applied, using the
193 '--cluster_unoise' command to denoise reads with default parameters (--unoise_alpha 2 --minsize 8)
194 while filtering for sequences without the expected length (--minseqlength 195 --maxseqlength 215), the
195 '--uchime3_denovo' command to remove chimeric sequences with default parameters, the
196 '--cluster_size' command to cluster at 99% similarity (--id 0.99), and the '--usearch_global' option to
197 map reads back to the retained OTUs (operational taxonomic units), using a 99% identity threshold (-
198 id.99) (Rognes et al. 2016). Next, LULU was used to merge similar OTUs (>84% similarity) with high
199 levels of co-occurrence (>95%; Frøslev et al. 2017). OTUs not matching the expected length (within
200 205 ± 3 bp; Vamos et al. 2017) were excluded. To account for potential contamination, the number of
201 reads from extraction and PCR negative controls was subtracted from associated samples.

202 The resulting OTUs were compared to the BOLD database using BOLDigger v2.1.1 (on 10–02–
203 2025; Buchner and Leese 2020) and identifications were manually curated. OTUs sequences were then
204 manually compared with BOLD hits, complemented with comparison with NCBI hits (in February
205 2025) and identified to the lowest possible taxonomic level based on the following thresholds: 98.5%
206 for species level, 97% for genus, 95% for family, and 90% for order. When OTUs had similar identities

207 across multiple species, genera, or other taxonomic ranks, the broadest level of classification was
208 selected, following a least common ancestor approach. Due to the incompleteness of COI databases for
209 West Africa, particularly on invertebrates and bats, OTUs not identifiable to species level were named
210 as the lowest taxonomic rank identified followed by their OTU number (e.g. Aranea OTU1021,
211 *Scotophilus* OTU2). OTUs were categorized as dietary (e.g., most arthropods) or non-dietary (e.g.,
212 vertebrates, fungi, internal and external parasites, OTUs only identified to class level).

213 It's important to acknowledge the limitations of DNA metabarcoding, particularly the potential for
214 detecting secondary ingestion. This can occur when prey species consumed by arthropod predators,
215 which are then eaten by bats and birds, are detected (da Silva et al. 2019; Deagle et al. 2019). However,
216 since the DNA from the prey gut contents is usually present in smaller quantities and more prone to
217 degradation, the likelihood of amplification during PCR is lower (Mata et al. 2021). Additional filtering
218 was applied to exclude samples containing fewer than 100 dietary reads and to discard diet OTUs
219 accounting for less than 1% of the total reads per sample, thereby minimizing the effects of secondary
220 ingestion (Deagle et al. 2019).

221 **Data analysis**

222 Predators represented by fewer than seven sampled individuals were excluded from the analysis. While
223 this sample size is below optimal levels for a detailed diet characterisation (Mata et al. 2019), this
224 threshold was chosen to retain a broader representation of bird species, particularly insectivorous
225 species, and to allow meaningful comparisons with bats.

226 Diet breadth and sample coverage for each predator species were assessed using the iNEXT function
227 from the iNEXT package (Hsieh et al. 2016). Rarefaction and sample coverage curves were generated
228 from sample-based incidence data using Hill number species richness ($q=0$), with 1000 bootstrap
229 replicates. Curves were calculated for three taxonomic prey levels (OTU, family, and order), along with
230 95% confidence intervals. To allow direct comparison of diet breadth among predators, we extracted
231 richness estimates standardised to 14 individuals. This value corresponds to twice the minimum number
232 of individuals sampled for any included predator species, including the species with the lowest sample

233 coverage, and represents a compromise between maximising species inclusion and ensuring comparable
234 sampling effort across predators (Chao et al. 2014).

235 Diet composition was analysed using occurrence-based metrics due to the high dietary richness and
236 potential recovery bias associated with differential digestion of prey (Deagle et al. 2019). The weighted
237 percent of occurrence (wPOO) was calculated following Deagle et al. (2019) at three prey taxonomic
238 levels (OTU, family, and order). This metric was calculated for all predator individuals combined, for
239 birds and bats separately, and for each predator species. OTUs that could not be assigned to family or
240 order were retained in the dietary matrices as missing values (NA) at those taxonomic levels.

241 Dietary differences between predators were visualised using non-metric multidimensional scaling
242 (NMDS) using the metaMDS function from the vegan package (Oksanen et al. 2022). Differences in
243 dietary composition were tested using analysis of similarities (ANOSIM) with the anosim function from
244 the same package, using 999 permutations. Species-level NMDS and ANOSIM analyses were based on
245 wPOO data and distances were calculated with the Bray-Curtis index, while individual-level analysis
246 used prey occurrence data and distances were calculated with the Jaccard index. For individual-level
247 analysis, OTUs found in only a single individual's diet were excluded. This approach reduces noise
248 associated with extremely rare prey while preserving individual-level dietary variation.

249 To further mitigate the influence of zero inflation associated with rare prey items, dietary overlap was
250 quantified using Pianka's index, which emphasises shared prey items by placing greater weight on their
251 presence rather than their absence (Pianka 1973). Pianka's index and its statistical significance were
252 computed using wPOO values and the niche_null_model function from the EcoSimR package (Gotelli
253 et al. 2015), with 1000 randomisations, and using the RA3 as randomisation algorithm. This algorithm
254 reshuffles the prey species wPOO within each predator, while retaining the observed diet breadth. For
255 each predator pair, analysis was restricted to prey taxa detected in the diet of at least one of the two
256 predators, thereby avoiding the zero inflation that would arise from a community-wide prey matrix.

257 Differences in prey consumption between bats and birds at the prey order level were assessed using
258 chi-square tests. Specifically, we compared the number of bat and bird individuals consuming each prey

259 order. Adjusted p-values were obtained using the Benjamini–Hochberg false discovery rate (FDR)
260 correction for multiple comparisons. Differences in predator and prey composition among rice fields
261 were assessed using ANOSIM with 999 permutations. Predator composition was analysed using
262 occurrence data and distances were calculated with the Jaccard index, while prey composition was based
263 on daily prey wPOO and distances were calculated with the Bray-Curtis index.

264 To examine the relationship between phylogeny and diet, the correlation between diet dissimilarities and
265 phylogenetic distances was assessed with a multiple regression on distance matrices (MRM) using the
266 MRM function from the ecodist package (Goslee and Urban 2007). For each predator group, 1000
267 phylogenetic trees were sampled. Bird trees were retrieved from BirdTree (Jetz et al. 2012) using the
268 tree source “Ericson All Species”, and bat trees were retrieved from VertLife (Upham et al. 2019) using
269 the tree source “Mammals birth-death node-dated completed trees”. For predators identified only to the
270 genus level, a congeneric species known to occur in the study area and with similar morphological traits
271 was used as a proxy. We then computed the average phylogenetic distance for each predator pair, based
272 on the 1000 trees. Diet dissimilarity was calculated for each predator pair as $D = 1 - O$, where D is the
273 dissimilarity index and O is Pianka’s index.

274 To account for variation in sampling effort, we included the minimum sample size of each pair of
275 predators in the regression. As a sensitivity analysis, we performed the MRM with different pools of
276 species, assessing the influence of phylogenetic distant species (e.g., *Ispidina picta*), and the influence
277 of species with very small and very large sample sizes for both bats and birds.

278 All graphs were generated using the ggplot function from the ggplot2 package (Wickham 2016), and all
279 analyses were performed in the R statistical environment (R Core Team 2024). Statistical significance
280 was set at $\alpha = 0.05$.

281 **Results**

282 We analysed the diets of 471 individuals, comprising 334 bats and 137 birds, across 13 species of bats
283 and eight species of birds. DNA metabarcoding of 665 faecal samples revealed 3371 occurrences of
284 1304 different arthropod prey OTUs, spanning 144 families and 22 orders. Only 10.1% of prey OTUs

285 are identified at the species level.

286 The average number of individuals sampled per predator species was 22,4 (± 10.3 , 95% CI). *Scotophilus*
287 sp. was the most sampled taxon, with 106 individuals (Table 1). In contrast, *Nycteris* cf. *hispidus*, *Prinia*
288 *subflava*, and *Scotoecus albofuscus* were each represented by seven individuals. All these species
289 exhibited sample coverage values above 22%, with the exception of *N. cf. hispidus*, which showed an
290 exceptionally low sample coverage (0.05%), indicating that its detected diet represents only a small
291 fraction of its total prey richness. Accordingly, results involving this species should be interpreted with
292 caution, particularly in analyses of dietary composition and overlap.

293 **Diet breadth and composition**

294 On average, each predator species consumed 99.7 ± 37.4 different OTUs, with values ranging from 28
295 in *P. subflava* and *Uraeginthus bengalus* to 369 prey OTUs in *Scotophilus* sp. (Table 1).

296 The most frequently consumed prey orders were Hemiptera, Coleoptera, Blattodea, Lepidoptera, and
297 Orthoptera (Figure 2; Online resource 1). When considering bats alone, the same orders dominated the
298 diet and showed similar relative importance. In contrast, bird diets were dominated by Lepidoptera, with
299 Aranea showing higher representation compared to bats (Figure 2; Online resource 1). At the family
300 level, Termitidae was the most frequently detected family. However, only 69.1% of prey OTUs could
301 be identified to family level (Online resource 2). The most frequently detected prey species was
302 *Macrotermes bellicosus*, a termite species (Online resource 3). The consumption of this and other highly
303 represented termite species, such as *Microtermes* sp. showed a seasonal peak in June and July.

304 With the exception of *N. cf. hispidus*, which exhibited an exceptionally low sample coverage ($0.05 \pm$
305 0.05%), rarefaction analysis revealed that sample coverage at the OTU level ranged from $15.0 \pm 14.2\%$
306 in *Lagonosticta senegala* to $70.2 \pm 3.0\%$ in *Scotophilus* sp. (Table 1). Using richness estimates
307 standardised to 14 individuals, nearly all predator species consumed more than 40 different prey OTUs
308 (Table 1 and Figure 3). We observed a high proportion of rare OTUs in predator diets: about 60.8% of
309 prey OTUs in bats and 72.5% in birds were detected in the diet of only a single individual.

310 Using diet breadth standardized at 14 samples, bats generally exhibited broader niche breadths than
311 birds at the OTU level (Figure 3). However, this pattern was not evident at higher prey taxonomic levels,
312 where confidence intervals overlapped extensively among most predator species (Figure 3).
313 Additionally, we found no significant differences in dietary breadth between granivorous and
314 insectivorous birds (Figure 3), although sample sizes were limited.

315 *Uraeginthus bengalus* showed the narrowest niche breadth at the OTU level (34.33 ± 11.25 prey OTUs)
316 with Blattodea (family Termitidae) comprising almost half of its diet (Figure 2) with *Odontotermes* sp.,
317 accounting for 26.7% of it. Despite consuming a limited number of OTUs, these prey spanned a diversity
318 of families and orders similar to those exploited by other predators (Figure 3). Among bats,
319 *Glauconycteris variegata* showed the narrowest niche breadth across all prey taxonomic levels (Figure
320 3), feeding primarily on Lepidoptera, Blattodea, Orthoptera, and Hemiptera (Figure 2). In contrast, *L.*
321 *senegala* showed the broadest niche breadth among birds at the OTU level, with a mean number of prey
322 taxa (74.54 ± 17.18) comparable to that of bats, although its confidence intervals were wide and
323 overlapped with those of other bird species. *Euplectes franciscanus* was the bird species consuming the
324 highest richness of prey orders and was the only predator detected consuming Ephemeroptera,
325 Zygentoma, and Gastropoda.

326 **Diet overlap and phylogenetic relationships**

327 We detected compositional differences among rice fields for both prey (ANOSIM $R = 0.34$, $p = 0.001$)
328 and predator assemblages (ANOSIM $R = 0.44$, $p = 0.001$) indicating moderate spatial structuring of
329 predator and prey assemblages across the study area. This spatial heterogeneity warrants caution when
330 interpreting dietary overlap, as differences among predators may partly reflect variation in prey
331 assemblages across fields, even though all dietary analyses integrate individuals sampled across multiple
332 fields.

333 Despite some overlap, dietary composition differed between bats and birds, as shown by NMDS (Figure
334 4), which was supported by ANOSIM analysis (ANOSIM $R = 0.61$, $p = 0.001$). Differences between
335 the two predator groups were also evident in the distribution of prey orders. When comparing prey order

336 consumption between bats and birds, Hygrophila, Aranea, Odonata, Hymenoptera, and Lepidoptera,
337 were significantly more consumed by birds, while Dermaptera, Psocodea, Coleoptera, Blattodea and
338 Hemiptera were mainly consumed by bats (Online resource 4).

339 At the individual level, NMDS revealed a substantial overlap among predator species; however,
340 ANOSIM indicated a significant, albeit weak, differentiation among species overall ($R_{\text{all}} = 0.20$, $p =$
341 0.001), within bats ($R_{\text{Bats}} = 0.19$, $p = 0.001$), and within birds ($R_{\text{Birds}} = 0.11$, $p = 0.001$). This apparent
342 overlap was likely influenced by the high proportion of rare OTUs, which generated extensive zero
343 inflation even after excluding prey detected in only a single individual predator.

344 When reducing the influence of joint absences using Pianka's index, overall dietary overlap among
345 predators was low but significant ($O_{\text{all}} = 0.11$, $p = 0.001$). Overlap was more pronounced among bat
346 species ($O_{\text{Bats}} = 0.23$, $p = 0.001$), with several pairs of bat species showing higher overlap than expected
347 by chance (Figure 5). In contrast, bird species showed significant dietary segregation ($O_{\text{Birds}} = 0.05$, $p =$
348 0.017), with most pairwise comparisons showing lower overlap than expected (Figure 5). Pairwise
349 dietary overlap among birds was comparable to that observed between bat and bird species. However, a
350 small number of bird species pairs within the same family (e.g., Ploceidae and Estrildidae) showed non-
351 significant Pianka's index values (Figure 5).

352 Pianka's index also revealed relatively high dietary overlap among a group of morphological and
353 phylogenetic similar pipistrelloid bats (*Neoromicia somalica*, *Pseudoromicia rendalli*, and *Afronycteris*
354 *nanus*) (Figure 5). However, this overlap was strongly influenced by the high consumption of seasonally
355 abundant termites (*M. bellicosus* and *Microtermes sp.*). When these prey were excluded, overlap values
356 decreased overall; however, only the overlap between *N. somalica* and *P. rendalli* was no longer
357 significant ($O_{\text{Neo som} - \text{Afr nan}} = 0.39$, $p_{\text{Neo som} - \text{Afr nan}} = 0.028$; $O_{\text{Neo som} - \text{Pse ren}} = 0.34$, $p_{\text{Neo som} - \text{Pse ren}} = 0.174$;
358 $O_{\text{Pse ren} - \text{Afr nan}} = 0.42$, $p_{\text{Pse ren} - \text{Afr nan}} = 0.008$). Conversely, two species from distinct suborders
359 *Macronycteris gigas* and *Mops condylurus*, showed the highest overlap ($O = 0.72$, $p < 0.001$) which
360 remained high even after excluding termites ($O = 0.63$, $p = 0.005$).

361 Despite the higher overlap among some closely related species, no significant correlation was detected
362 between phylogenetic distance and diet dissimilarity at OTU prey level, for both predator groups (R^2_{Bats}
363 = 0.04, $p_{\text{Bats}} = 0.28$; $R^2_{\text{Birds}} = 0.06$, $p_{\text{Birds}} = 0.31$). However, for birds, when removing the *I. picta* from
364 the analysis, this relation became significant ($\beta = 0.0009$; $R^2 = 0.35$; $p = 0.01$). For birds, the minimum
365 sample size in each predator pair explained part of dietary variation ($\beta = -0.01$; $R^2 = 0.21$; $p = 0.02$),
366 with lower dissimilarity (higher overlaps) being correlated with higher minimum sample sizes.

367 **Discussion**

368 In this study, we applied DNA metabarcoding to examine the dietary breadth, composition, and trophic
369 overlap in a community of invertebrate-eating bats and birds foraging in lowland rainfed rice fields in
370 West Africa. By integrating diet data across multiple predator species and taxonomic groups, our results
371 provide one of the first metabarcoding-based community-level assessments of trophic resource use by
372 flying vertebrates in African agricultural landscapes. Overall, we found broad and relatively generalised
373 diets across predators, differences in dietary composition between bats and birds, reduced dietary
374 overlap among birds, and moderate overlap among some bat species. These patterns emerge from a
375 spatially heterogeneous landscape in which both prey and predator assemblages vary among rice fields
376 and therefore represent integrated trophic relationships across multiple foraging contexts rather than
377 field-specific interactions.

378 **Diet breadth and prey use**

379 Both bats and birds displayed broad diets characterised by high prey richness and a large proportion of
380 rare prey items. This pattern is consistent with expectations for tropical systems, where high arthropod
381 diversity and strong temporal variation in prey availability thought to favour generalised foraging
382 patterns (Novotny and Miller 2014). Bats generally showed broader dietary breadths than birds at the
383 OTU level, in line with previous studies suggesting that nocturnal insectivorous predators often exploit
384 a wider range of prey taxa than diurnal predators (Maas et al. 2016; Wong and Didham 2024). This
385 difference may reflect a combination of higher nocturnal insect activity in warmer environments,
386 reduced visual constraints, and flexible foraging strategies in bats (Boonman et al. 2013; Kirkpatrick et

387 al. 2018; Wong and Didham 2024). At coarser taxonomic levels (family and order), however, niche
388 breadth estimates converged across predators, with extensive overlap in confidence intervals. This
389 indicates that differences between bats and birds are driven primarily by fine-scale prey selection rather
390 than by the exploitation of fundamentally different prey groups (Heim et al. 2021; Chaves et al. 2026).
391 Similarly, we detected no significant differences in dietary breadth between granivorous and
392 insectivorous birds, although this result should be interpreted cautiously given the limited sample sizes
393 for some insectivorous species.

394 A high proportion of rare OTUs (prey items recorded only once) was detected in both bat and bird diets.
395 These results suggest that some bird species may exhibit broader diets than often assumed, while
396 reinforcing the generally broad dietary character of bats (Maas et al. 2016). Although consistent with
397 previous metabarcoding studies (Garfinkel et al. 2022; Bookwalter et al. 2023), the high number of rare
398 OTUs may also partly reflect unequal and, in some cases, low sample coverage, which can bias the
399 detection of infrequently consumed prey, particularly in birds. Rare prey items should thus be interpreted
400 cautiously; nevertheless, their prevalence, together with the consumption of prey during seasonal peaks
401 is consistent with, and may reflect, opportunistic foraging behaviour reported in other systems (Mwansat
402 et al. 2015; Andriollo et al. 2021).

403 The general patterns of prey order importance for bats and birds were consistent with general knowledge
404 for other world regions (Kunz et al. 2011; Nyffeler et al. 2018). Among the most frequently detected
405 prey items, termites (Blattodea; Termitidae) and ants (Hymenoptera; Formicidae) emerged as major
406 components of bats' and birds' diets. Ant and termite abdomens, rich in fat (Eklöf and Rydell 2017), are
407 a critical food source in the tropics for many animals, including bats and birds (Redford 1987; Korb and
408 Salewski 2001). For example, *U. bengalus* showed the highest termite consumption, consistent with
409 previous observations that termites are vital to the diet of nestlings of this species (Hamed and Evans
410 1983). Other predators such as *G. variegata*, and *P. rendalli*, consumed termites seasonally (e.g. *M.*
411 *bellicosus* and *Microtermes sp.*), exhibiting temporal shifts in diet corresponding to known termite
412 reproductive periods, typically peaking between late June and early July when alate termites emerge
413 shortly after the first heavy rains (Mitchell 2008; Manzoor and Mir 2010).

414 The marked contribution of termite taxa to predator diets suggests that temporally abundant prey can
415 strongly influence observed dietary patterns, including increasing apparent overlap among species.
416 Chaves et al. 2026 also reported the consumption of agricultural pests by these predators foraging in
417 rice fields. The combination of broad diets and the consumption of pest taxa indicates a potential
418 contribution of these predators in pest suppression, including rice pests, which often exhibit periodic
419 peaks of abundance similar to termites (Heinrichs 1994). This potential role cannot be directly quantified
420 from dietary data alone; however it suggests that these predators may contribute to pest management in
421 West African rice fields, as already documented for other regions and cropping systems (Díaz-Sieffer et
422 al. 2022; Tuneu-Corral et al. 2023).

423 However, several limitations constrain inference on predator diet and pest control potential, including
424 incomplete diet characterisation, as evidenced by low sample coverage for some species, incomplete
425 reference databases, which resulted in a low proportion of OTUs identified to the species level (10,1%),
426 and limited ecological knowledge of many invertebrate taxa in the region. Despite these limitations, the
427 frequent detection of diverse arthropod taxa, including pest species (Chaves et al. 2026), highlights the
428 ecological relevance of bats and birds within these agroecosystems.

429 Overall, the broad diets of flying vertebrates in West African rice fields likely reflect a combination of
430 diverse prey assemblages and the flexible foraging strategies (MacArthur and Pianka 1966; Pyke 2019).
431 Although both groups exhibited broad dietary patterns, some species showed more restricted prey use,
432 suggesting some degree of trophic differentiation. These dietary differences likely stem from a
433 combination of behavioural, morphological, and temporal factors, although their relative importance
434 cannot be disentangled with the present data

435 **Diet overlap and phylogenetic relationships**

436 We observed differences in dietary composition between bats and birds, consistent with differences in
437 foraging period and prey activity patterns. Diurnal birds consumed a higher proportion of typically
438 diurnal prey orders such as Odonata and Hymenoptera (Corbet 1980; Wong and Didham 2024), whereas
439 bats primarily consumed nocturnal or crepuscular taxa such as Dermaptera, Coleoptera, Blattodea, and

440 Hemiptera (Rankin and Palmer 2009; Wong and Didham 2024). Therefore, temporal segregation likely
441 represents an important axis of niche differentiation between these two predator groups.

442 Within groups, patterns of dietary overlap differed markedly between bats and birds. Bird species
443 exhibited low dietary overlap, with most species' pairs showing lower overlap than expected by chance.
444 Only a small number of within-family comparisons yielded non-significant Pianka's index values,
445 indicating neither overlap nor clear segregation. Despite the very low overlap values, the analysis of
446 correlation between diet dissimilarity and phylogenetic distances revealed a weak relationship only after
447 excluding *I. picta*, a phylogenetically distant species. This suggests that phylogeny may contribute to
448 dietary dissimilarity, although this result is sensitive to species composition. However, dietary
449 dissimilarity is also influenced by sample sizes. Therefore, these results may reflect a combination of
450 ecological differences and sampling effects, rather than clear phylogenetic structuring (Deagle et al.
451 2019; Mata et al. 2019). Overall, the relationship between phylogeny and diet remains inconclusive in
452 this system.

453 The low overlap among bird species may also reflect a greater diversity in foraging strategies, as this
454 study included exclusively insectivorous, primarily granivorous, and granivorous-insectivorous bird
455 species (Owiunji and Plumptre 1998; Billerman et al. 2022). Even among exclusively insectivorous, we
456 examined highly aerial foragers, such as *H. lucida* (Turner and Rose 1989), and ground-foraging species
457 such as *P. subflava* and *I. picta* (Ryan and Dean 2020; Woodall 2021).

458 Bat species showed higher dietary overlap than birds, particularly among morphologically and
459 phylogenetically similar pipistrelloid species. This overlap was also partially driven by the high
460 consumption of termites (*M. bellicosus* and *Microtermes* sp.). The high importance of these prey species,
461 as measured by wPOO, is influenced by the feeding behaviour of individuals exploiting swarming
462 termites or other temporally abundant prey, which can inflate their apparent contribution to the overall
463 diet. Although these termites were a significant component of the bat diet, the relatively high dietary
464 overlap observed between bats, particularly between closely related species, does not necessarily
465 indicate low competition, as overlap in resource use does not directly translate into competitive
466 interactions (Schoener 1974). Instead, this pattern may reflect shared exploitation of abundant prey

467 resources and similar foraging strategies (Findley and Black 1983; Schoeman and Jacobs 2011;
468 Dammhahn and Goodman 2014; Blanch et al. 2023). Several factors may contribute to this pattern,
469 including the high mobility and extensive home ranges of bats (Fenton 1990), and variation in habitat
470 use and resource distribution (Stevens and Willig 2000). Competition for food resources might only be
471 evident when food is scarcer (Roeleke et al. 2018) probably during the dry season. Population regulation
472 from predation or disturbances may also contribute to reduced density, reducing competition (Bloch et
473 al. 2011). However, the relative importance of these mechanisms cannot be determined from the present
474 data.

475 Despite the higher overlap among some closely related bat species, no significant relationship was
476 detected between phylogenetic distance and diet dissimilarity. The lack of correlation likely reflects a
477 combination of limited statistical power, diet variability, and potential uncertainty in phylogenetic
478 relationships (López-Aguirre et al. 2023). The high overlap between *Macronycteris gigas* and *Mops*
479 *condylurus*, two species from distinct suborders, further illustrates that dietary similarity can occur
480 among distantly related taxa.

481 Overall, while temporal foraging behaviour plays a key role in structuring dietary differences between
482 bats and birds, our findings suggest that niche partitioning is a more complex process influenced by
483 multiple factors, including morphology, behaviour, and habitat use. These factors, alongside foraging
484 time, likely shape the dietary overlap observed within and between these predator groups.

485 **Conclusions**

486 Our findings highlight variation in diet breadth and overlap among invertebrate-eating flying vertebrates
487 foraging in lowland rainfed rice fields. The observed differences in dietary composition, shaped by
488 dietary preferences and foraging strategies, underscore the importance of preserving diverse
489 communities of birds and bats within these agroecosystems. At the same time, dietary overlap among
490 some bat species suggests shared use of available prey resources. Moreover, the generalised diets of
491 these predators, together with the consumption of pest taxa, indicate a potential role in the regulation of
492 invertebrate populations. The critical importance of rice in global food security (Maclean et al. 2002),

493 and the harm that chemical pesticides pose to human health and wildlife (Ansari et al. 2014), highlight
494 the need to better understand the ecological roles of these invertebrate-eating predators in rice systems.
495 Key measures to foster the presence of birds and bats in agroecosystems, while aligning with sustainable
496 agricultural practices, might include maintaining patches of native vegetation (Horgan et al. 2017) and
497 isolated trees within the rice fields (Fernandes 2024), providing permanent drinking water sources
498 during the dry season, ensuring adequate roosting sites, and reducing agrochemical usage (Tuneu-Corral
499 et al. 2023). However, some of these predators can also affect crops directly (de Mey et al. 2012; Downs
500 and Hart 2020) or influence other trophic interactions, including mesopredator regulation (Sottomayor
501 et al. 2024). Therefore, further studies are needed to provide a deeper understanding of the true
502 ecological roles of these bats and birds.

503 Given the high environmental variability across seasons, further studies focusing on temporal variation
504 in dietary composition and overlap would help to better characterize the trophic structure of these
505 predator communities in West African agroecosystems. Advancing this understanding will be essential
506 for integrating ecological knowledge into the sustainable management of tropical agroecosystems.

507

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518 **Conflicts of interest**

519 The authors declare that they have no conflict of interest.

520 **Ethics approval**

521 All applicable institutional and/or national guidelines for the care and use of animals were followed.

522 **Availability of data and material**

523 The datasets used and analysed during the current study are available from the corresponding author on
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792 Tables

793 Table 1 – Predator species and prey diversity. The table lists predator species grouped into bats and birds,
794 showing sample size (the number of individual predators analysed), the total number of different prey
795 taxa (OTU) found, the number of prey found in the diet of 14 individuals (mean \pm 95% CI), and the
796 sample coverage for each species (mean \pm 95% CI). Abbreviations used for species names are in the
797 "Acronym" column.

Predator species	Acronym	Group	Sample size	No of prey OTUs	No of prey OTUs for 14 individuals	Sample Coverage
<i>Afronycteris nanus</i>	<i>Afr nan</i>	Bat	36	126	61.11 \pm 7.28	0.60 \pm 0.06
<i>Euplectes franciscanus</i>	<i>Eup fra</i>	Bird	13	58	61.95 \pm 14.29	0.22 \pm 0.14
<i>Glauconycteris variegata</i>	<i>Gla var</i>	Bat	11	38	44.48 \pm 9.34	0.54 \pm 0.16
<i>Hipposideros caffer/ ruber</i>	<i>Hip caf/rub</i>	Bat	34	242	120.60 \pm 10.59	0.50 \pm 0.06
<i>Hirundo lucida</i>	<i>Hir luc</i>	Bird	8	31	49.18 \pm 15.06	0.27 \pm 0.20
<i>Ispidina picta</i>	<i>Isp pic</i>	Bird	9	39	55.08 \pm 13.83	0.30 \pm 0.20
<i>Lagonosticta senegala</i>	<i>Lag sen</i>	Bird	10	55	74.54 \pm 17.18	0.15 \pm 0.14
<i>Macronycteris gigas</i>	<i>Mac gig</i>	Bat	10	52	68.34 \pm 14.84	0.36 \pm 0.14
<i>Mops condylurus</i>	<i>Mop con</i>	Bat	41	185	82.98 \pm 7.65	0.62 \pm 0.05
<i>Neoromicia somalica</i>	<i>Neo som</i>	Bat	23	106	73.36 \pm 9.77	0.56 \pm 0.08
<i>Nycteris cf. hispida</i>	<i>Nyc cf his</i>	Bat	7	49	95.69 \pm 25.67	0.05 \pm 0.05
<i>Pipistrellus</i> sp.	<i>Pip sp</i>	Bat	13	78	82.21 \pm 12.26	0.52 \pm 0.10
<i>Ploceus cucullatus</i>	<i>Plo cuc</i>	Bird	71	202	53.75 \pm 5.33	0.53 \pm 0.06
<i>Ploceus melanocephalus</i>	<i>Plo mel</i>	Bird	8	29	45.69 \pm 13.85	0.31 \pm 0.20
<i>Prinia subflava</i>	<i>Pri sub</i>	Bird	7	28	47.45 \pm 14.55	0.29 \pm 0.23
<i>Pseudoromicia rendalli</i>	<i>Pse ren</i>	Bat	18	128	104.28 \pm 13.99	0.41 \pm 0.08
<i>Scotoecus albofuscus</i>	<i>Scoe alb</i>	Bat	7	72	123.59 \pm 25.69	0.33 \pm 0.13
<i>Scotoecus</i> sp.	<i>Scoe sp</i>	Bat	19	128	103.21 \pm 12.41	0.59 \pm 0.07
<i>Scotophilus leucogaster</i>	<i>Scop leu</i>	Bat	9	51	73.49 \pm 16.69	0.30 \pm 0.15
<i>Scotophilus</i> sp.	<i>Scop sp</i>	Bat	106	369	83.51 \pm 5.19	0.70 \pm 0.03
<i>Uraeginthus bengalus</i>	<i>Ura ben</i>	Bird	11	28	34.33 \pm 11.25	0.33 \pm 0.19

798

799 Figure legends

800 Figure 1 – Location of the study site in Guinea-Bissau. The map on the left shows the location of
801 sampling points and respective rice fields. The map was built under the coordinate reference system
802 EPSG:4326 – WGS 84 on 22 November 2024. Insets include a map with the location of the study area
803 in the context of West Africa and the location of Oio Region and Mansabá Sector in Northern Guinea-
804 Bissau (top-right). The pictures on the right show typical lowland rainfed smallholder rice fields, namely
805 Demba Só (centre) and Mambonco (bottom).

806 Figure 2 – Weighted percentage of occurrence (wPOO) of arthropod orders in the diet of bats and birds
807 from lowland rainfed rice fields of Guinea Bissau. The first column on the left represents the combined
808 diet of all species, while the second and third columns represent the diets of all bat species combined
809 and all bird species combined, respectively. The following columns represent the diet of each species
810 with bats labelled in red, and birds labelled in blue. Predator species acronyms are listed in Table 1.

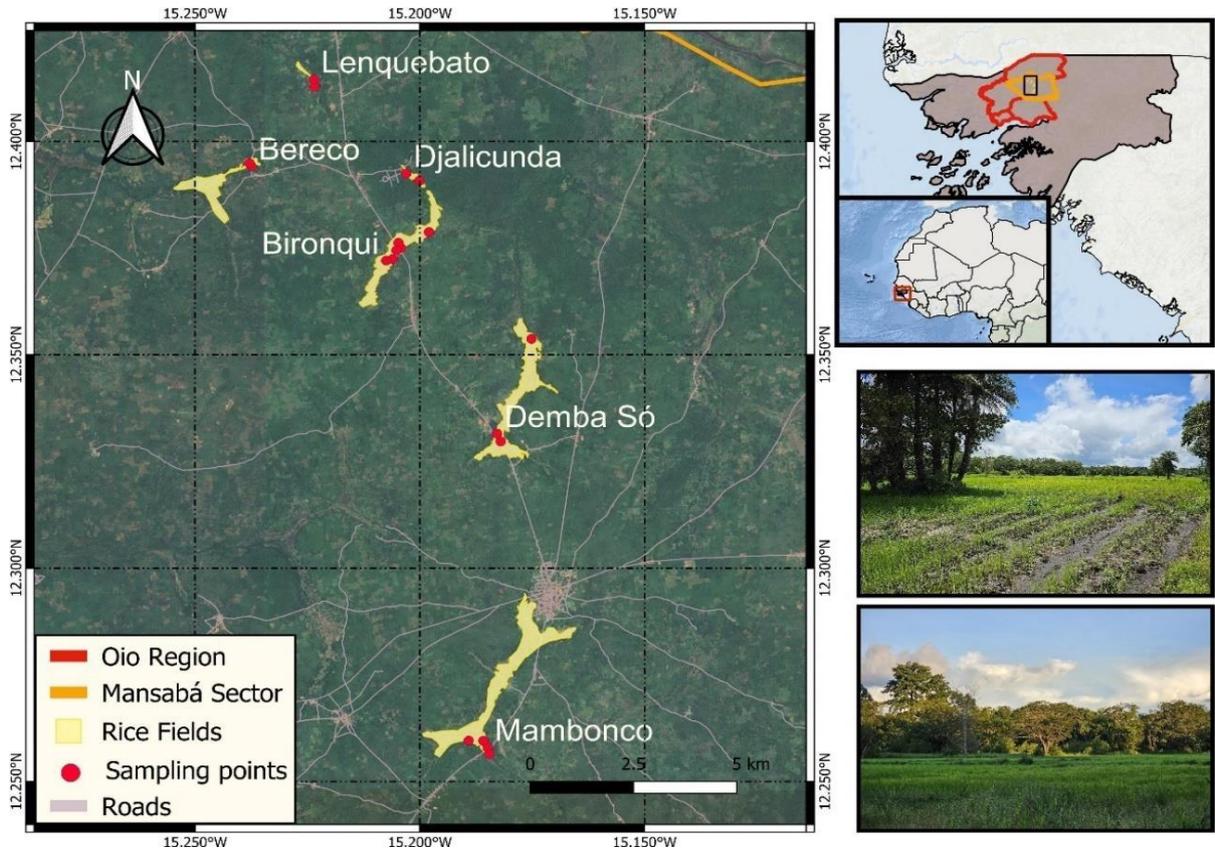
811 Figure 3 – Niche breadth estimates (taxon richness) for the 21 bat and bird species from lowland rainfed
812 rice fields of Guinea Bissau. Estimates are represented by black dots and coloured bars represent the
813 95% lower and upper confidence intervals. Niche breadth was estimated at different prey taxonomic
814 levels (OTU, Family, Order) for each species. Estimates were calculated for 14 individuals and were
815 generated from sample-based incidence data, using Hill number species richness ($q=0$), and 1000
816 bootstraps. Bats bars are in red, while birds are in blue. Predator species acronyms are listed in Table 1.

817 Figure 4 – NMDS ordination of molecular bat and bird diets (at the OTU prey level) from lowland
818 rainfed rice fields of Guinea Bissau. The diamonds represent predator species, bats are labelled in red,
819 while birds are labelled in blue. Polygons represent the convex hulls enclosing 95% of data points.
820 Smaller dots represent prey species coloured by their respective order, and larger dots are the centroids
821 of each prey order. The ordination is calculated with Jaccard distance. The ordination stress level is
822 indicated in the panel. Predator species acronyms are listed in Table 1.

823 Figure 5 – Pianka's pairwise overlap of molecular bat and bird diets (at the OTU level) from lowland
824 rainfed rice fields of Guinea Bissau. The pair comparisons that show higher overlap values than expected
825 by chance are coloured in green and the ones that show lower values than expected by chance are

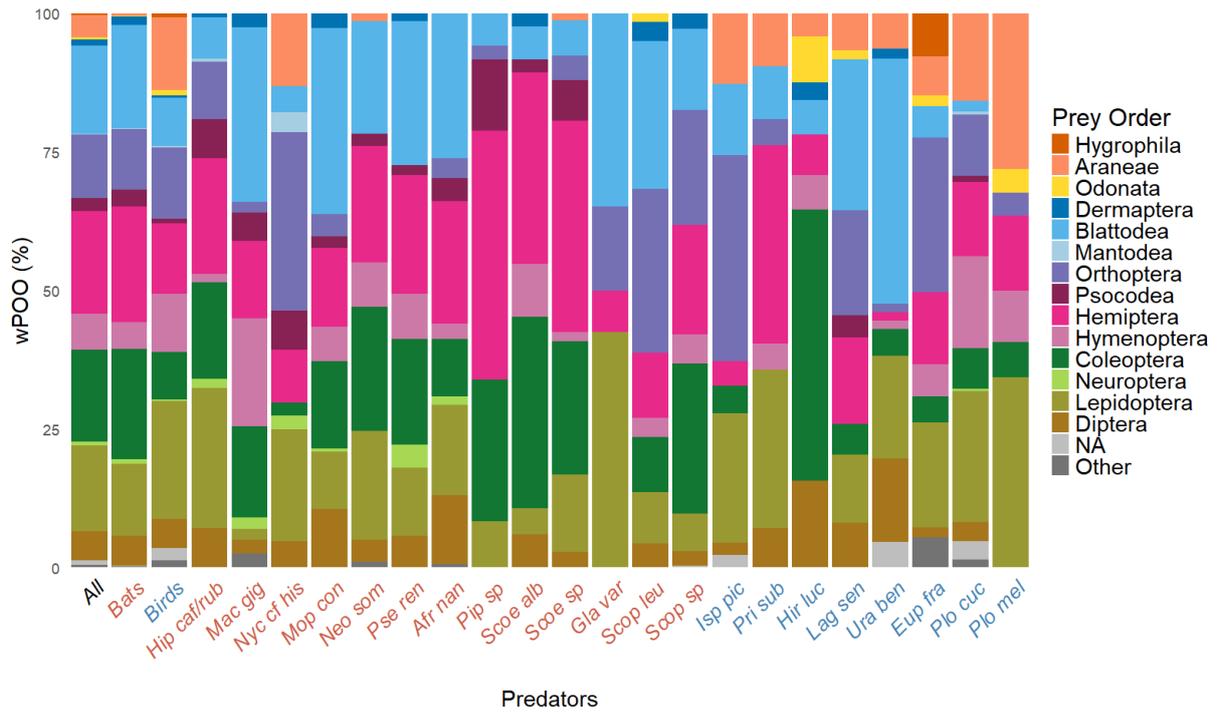
826 coloured in red. Values close to expectations are coloured in grey. Bats are labelled in red, while birds
827 are labelled in blue. Predator species acronyms are listed in Table 1.

828 Figures



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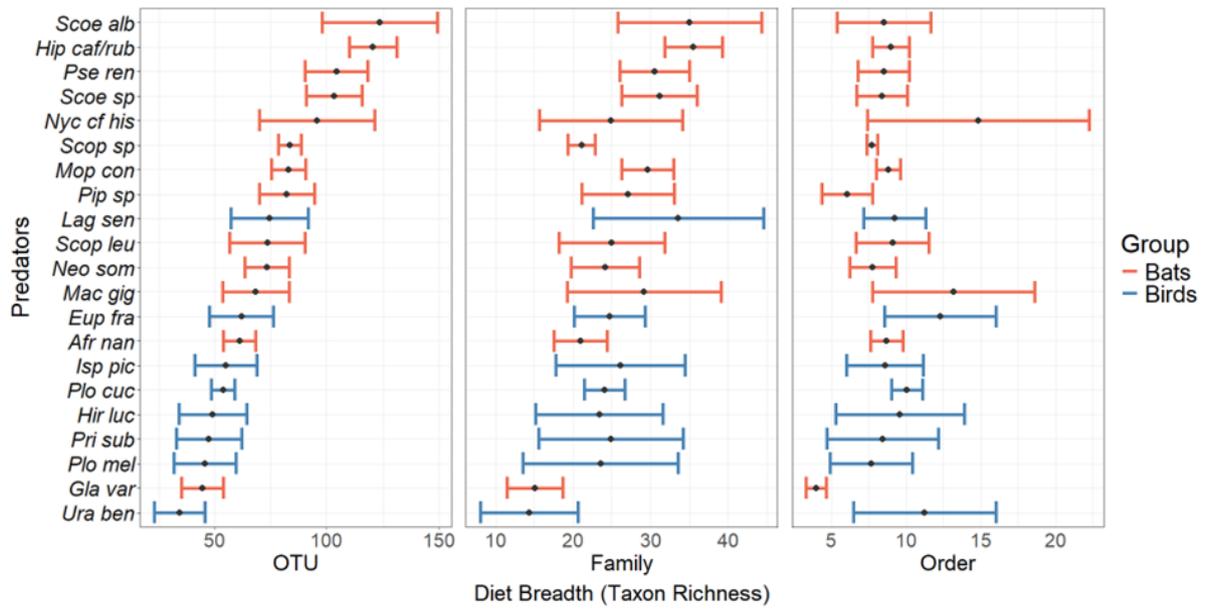
830 Figure 1



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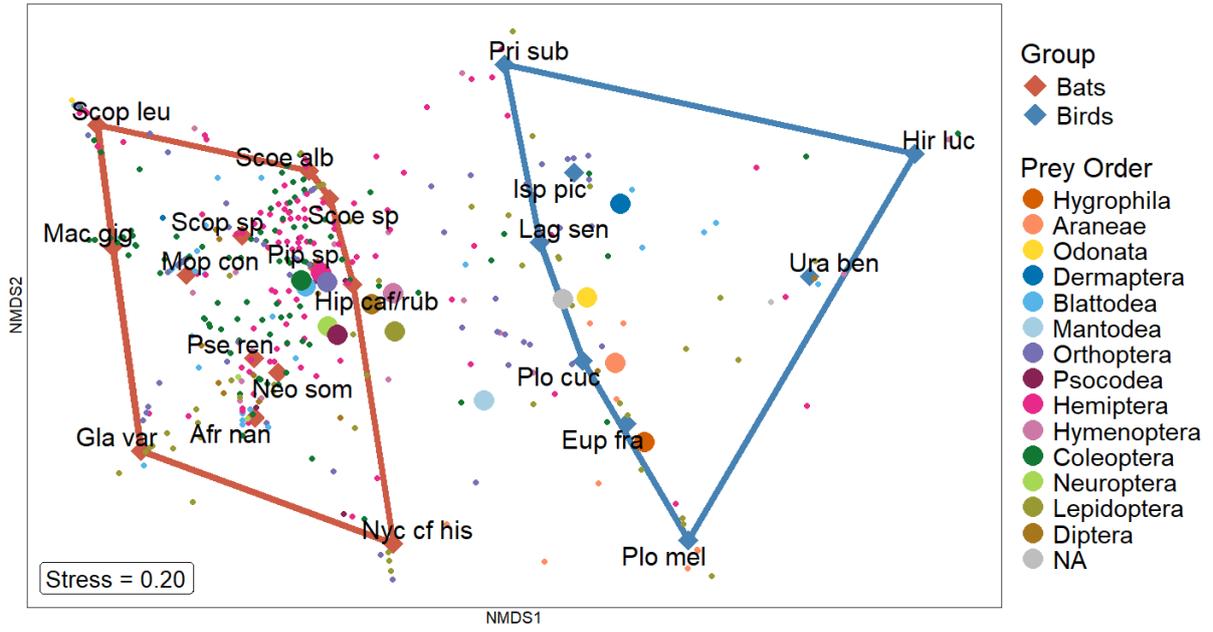
Predators

832 Figure 2



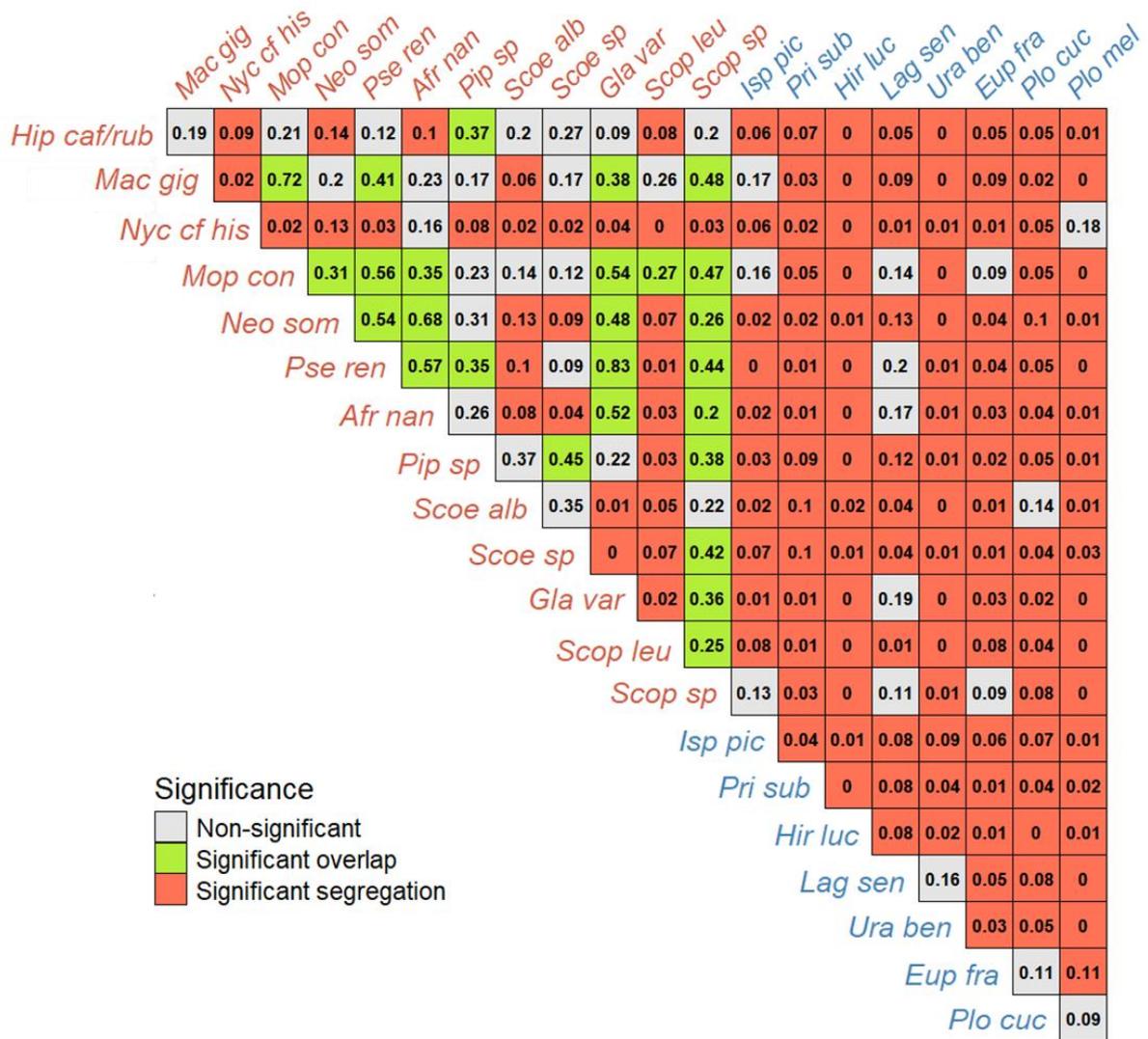
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834 Figure 3



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836 Figure 4



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838 Figure 5

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846 **Online Resources**

847 Online resource 1 – Weight percentage of occurrence (wPOO) of the prey orders consumed by all
 848 individual predators, and by bats and birds separately.

Prey Orders	wPOO (%)
All	
Hemiptera	18,58
Coleoptera	16,62
Blattodea	15,84
Lepidoptera	15,51
Orthoptera	11,45
Diptera	6,45
Hymenoptera	5,34
Araneae	4,12
Psocodea	2,34
Dermaptera	1,20
NA	0,71
Neuroptera	0,59
Odonata	0,33
Hygrophila	0,21
Mantodea	0,19
Ephemeroptera	0,11
Trichoptera	0,11
Decapoda	0,10
Lithobiomorpha	0,07
Phasmatodea	0,05
Scolopendromorpha	0,04
Zygentoma	0,04
Bats	
Hemiptera	20,96
Coleoptera	19,89
Blattodea	18,75
Lepidoptera	13,10
Orthoptera	10,91
Diptera	5,37
Hymenoptera	4,82
Psocodea	2,93
Dermaptera	1,56
Neuroptera	0,76
Araneae	0,45
Trichoptera	0,15
Mantodea	0,13
NA	0,10
Scolopendromorpha	0,06
Odonata	0,04
Birds	
Lepidoptera	21,39
Araneae	13,07
Hemiptera	12,77
Orthoptera	12,74

Hymenoptera	10,41
Blattodea	8,75
Coleoptera	8,64
Diptera	5,25
NA	2,20
Odonata	1,03
Psocodea	0,90
Hygrophila	0,73
Ephemeroptera	0,36
Decapoda	0,33
Dermaptera	0,33
Mantodea	0,33
Lithobiomorpha	0,24
Neuroptera	0,18
Phasmatodea	0,18
Zygentoma	0,15

849

850 Online resource 2 – Weight percentage of occurrence (wPOO) of the main prey families (over 1% of

851 wPOO) consumed by all individual predators, and by bats and birds separately.

Prey Families	wPOO (%)
All	
NA	18,19
Termitidae	8,20
Carabidae	5,23
Formicidae	4,57
Ectobiidae	4,47
Cicadellidae	4,06
Gryllidae	3,50
Rhyparochromidae	3,12
Erebidae	2,70
Pentatomidae	2,57
Tettigoniidae	2,23
Acrididae	2,18
Hydrophilidae	1,66
Gelechiidae	1,62
Scarabaeidae	1,47
Gracillariidae	1,43
Araneidae	1,41
Noctuidae	1,31
Delphacidae	1,27
Cydnidae	1,23
Chrysomelidae	1,16
Staphylinidae	1,16
Myopsocidae	1,15
Trigonidiidae	1,09
Limoniidae	1,03
Cicadidae	1,02
Bats	
NA	17,20

Termitidae	8,65
Carabidae	7,32
Ectobiidae	5,78
Cicadellidae	4,97
Gryllidae	4,29
Rhyparochromidae	3,99
Formicidae	3,25
Pentatomidae	2,88
Hydrophilidae	1,90
Acrididae	1,83
Scarabaeidae	1,76
Erebidae	1,72
Cydnidae	1,69
Delphacidae	1,66
Tettigoniidae	1,62
Gelechiidae	1,53
Staphylinidae	1,50
Trigonidiidae	1,46
Myopsocidae	1,43
Limoniidae	1,40
Nitidulidae	1,29
Chrysomelidae	1,12
Elateridae	1,08
Birds	
NA	20,60
Formicidae	7,80
Termitidae	7,11
Erebidae	5,07
Araneidae	4,16
Gracillariidae	3,93
Tettigoniidae	3,70
Cicadidae	3,50
Acrididae	3,03
Noctuidae	2,64
Cicadellidae	1,85
Gelechiidae	1,83
Pentatomidae	1,80
Gryllidae	1,58
Pisauridae	1,40
Salticidae	1,33
Geometridae	1,31
Ectobiidae	1,28
Chrysomelidae	1,26
Curculionidae	1,21
Ceratopogonidae	1,16
Hydrophilidae	1,06
Notodontidae	1,05
Rhyparochromidae	1,00

- 853 Online resource 3 – Weight percentage of occurrence (wPOO) of the main prey species (over 1% of
 854 wPOO) consumed by all individual predators, and by bats and birds separately.

Prey Species	wPOO (%)
All	
<i>Macrotermes bellicosus</i>	4.69
<i>Ectobiidae sp.</i>	2.55
<i>Oecophylla longinoda</i>	1.57
<i>Paraphonus</i>	1.52
<i>Microtermes sp.</i>	1.29
<i>Platymestopus vestitus</i>	1.25
<i>Pentatomidae sp.</i>	1.02
Bats	
<i>Macrotermes bellicosus</i>	6,46
<i>Ectobiidae sp.</i>	3,39
<i>Paraphonus sp.</i>	2,15
<i>Microtermes sp.</i>	1,78
<i>Platymetopus vestitus</i>	1,76
<i>Pentatomidae sp.</i>	1,29
<i>Cicadellidae sp.</i>	1,12
<i>Lepidoptera sp.</i>	1,06
Birds	
<i>Oecophylla longinoda</i>	4,47
<i>Cicadidae sp.</i>	3,26
<i>Erebidae sp.</i>	2,17
<i>Odontotermes sp.</i>	1,95
<i>Gracillariidae sp.</i>	1,67
<i>Lepidoptera sp.</i>	1,58
<i>Gelechiidae sp.</i>	1,57
<i>Odontotermes latericius</i>	1,36
<i>Diplopoda sp.</i>	1,36
<i>Orthoptera sp.</i>	1,10
<i>Lepidoptera sp.</i>	1,08
<i>Ancistrotermes cavithorax</i>	1,05
<i>Odontotermes nilensis</i>	1,03

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856

857 Online resource 4 – Distribution of prey orders consumed by predators, showing the total number of predator
 858 individuals analysed and the number and percentage of bird and bat individuals consuming each prey order.
 859 Adjusted p-values were obtained from chi-squared tests assessing differences between bird and bat frequencies
 860 for each prey order, with correction for multiple comparisons using the Benjamini–Hochberg false discovery
 861 rate (FDR).

Prey Orders	Number of predator individuals	Number of bird individuals	Number of bat individuals	Percentage of Birds	Percentage of bats	Adjusted p-value
Araneae	57	52	5	96,2	3,8	6,9E-27
Blattodea	192	29	163	30,3	69,7	2,2E-07
Coleoptera	238	34	204	28,9	71,1	9,7E-12
Decapoda	2	2	0	100,0	0,0	1,5E-01
Dermaptera	25	2	23	17,5	82,5	4,7E-02
Diptera	86	22	64	45,6	54,4	5,0E-01
Ephemeroptera	1	1	0	100,0	0,0	3,6E-01
Hemiptera	258	52	206	38,1	61,9	1,4E-05
Hygrophila	3	3	0	100,0	0,0	4,8E-02
Hymenoptera	95	38	57	61,9	38,1	2,5E-02
Lepidoptera	207	75	132	58,1	41,9	9,2E-03
Lithobiomorpha	1	1	0	100,0	0,0	3,6E-01
Mantodea	4	2	2	70,9	29,1	6,4E-01
NA	11	10	1	96,1	3,9	1,2E-04
Neuroptera	11	1	10	19,6	80,4	3,2E-01
Odonata	6	5	1	92,4	7,6	2,5E-02
Orthoptera	141	46	95	54,1	45,9	3,6E-01
Phasmatodea	1	1	0	100,0	0,0	3,6E-01
Psocodea	34	4	30	24,5	75,5	4,6E-02
Scolopendromorpha	1	0	1	0,0	100,0	1,0E+00
Trichoptera	2	0	2	0,0	100,0	1,0E+00
Zygentoma	1	1	0	100,0	0,0	3,6E-01

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