

1 **Testing ecological hypotheses at the pondscape with**
2 **environmental DNA metabarcoding: a case study on a**
3 **threatened amphibian**

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21

22 **Running title:** Hypothesis testing with eDNA metabarcoding

23

24 **Abstract**

25

26 Environmental DNA (eDNA) metabarcoding is revolutionising biodiversity monitoring, but has unrealised
27 potential for ecological hypothesis testing. Here, we validate this potential in a large-scale analysis of
28 vertebrate community data generated by eDNA metabarcoding of 532 UK ponds. We test biotic associations
29 between the threatened great crested newt (*Triturus cristatus*) and other vertebrates as well as abiotic factors
30 influencing *T. cristatus* occupancy at the pondscape. Furthermore, we test the status of *T. cristatus* as an
31 umbrella species for pond conservation by assessing whether vertebrate species richness is greater in ponds
32 with *T. cristatus* and higher *T. cristatus* Habitat Suitability Index (HSI) scores. *T. cristatus* occupancy was
33 positively correlated with amphibian and waterfowl species richness. Specifically, *T. cristatus* was positively
34 associated with smooth newt (*Lissotriton vulgaris*), common coot (*Fulica atra*), and common moorhen
35 (*Gallinula chloropus*), but negatively associated with common toad (*Bufo bufo*). *T. cristatus* occupancy did not
36 significantly decrease as fish species richness increased, but negative associations with common carp (*Cyprinus*
37 *carpio*), three-spined stickleback (*Gasterosteus aculeatus*) and ninespine stickleback (*Pungitius pungitius*) were
38 identified. *T. cristatus* occupancy was negatively correlated with mammal species richness, and *T. cristatus* was
39 negatively associated with grey squirrel (*Sciurus carolinensis*). *T. cristatus* occupancy was negatively influenced
40 by larger pond area, presence of inflow, and higher percentage of shading, but positively correlated with HSI
41 score, supporting its application to *T. cristatus* survey. Vertebrate species richness was significantly higher in *T.*
42 *cristatus* ponds and broadly increased as *T. cristatus* HSI scores increased. We reaffirm reported associations
43 (e.g. *T. cristatus* preference for smaller ponds) but also provide novel insights, including a negative effect of
44 pond inflow on *T. cristatus*. Our findings demonstrate the prospects of eDNA metabarcoding for ecological
45 hypothesis testing at landscape scale and dramatic enhancement of freshwater conservation, management,
46 monitoring and research.

47

48 **Keywords:** biodiversity assessment, environmental DNA (eDNA), hypothesis testing,
49 metabarcoding, ponds, species associations, *Triturus cristatus*

50

51 **1. Introduction**

52

53 Environmental DNA (eDNA) analysis offers ecologists exceptional power to detect organisms
54 within and across ecosystems. DNA released by organisms into their environment via
55 secretions, excretions, gametes, blood, or decomposition, can be sampled and analysed
56 using different approaches to reveal the distribution of single or multiple species (Rees et
57 al., 2014; Lawson Handley, 2015). eDNA analysis combined with high-throughput
58 sequencing (i.e. eDNA metabarcoding) can yield efficient, comprehensive assessments of
59 entire communities (Deiner et al., 2017), providing a step change in biodiversity monitoring
60 (Hering et al., 2018). eDNA metabarcoding has untapped potential to test ecological
61 hypotheses by enabling biodiversity monitoring at landscape scale with minimal impact to
62 communities under investigation. This potential has already been demonstrated with
63 targeted eDNA analysis by Wilcox et al. (2018), where climate-mediated responses of bull
64 trout (*Salvelinus confluentus*) to biotic and abiotic factors were revealed using quantitative
65 PCR (qPCR) on crowd-sourced eDNA samples. Although eDNA metabarcoding assessments
66 of alpha and beta diversity along environmental gradients are increasing (e.g. Hänfling et al.,
67 2016; Olds et al., 2016; Kelly et al., 2016; Evans et al., 2017; Li et al., 2018a; Nakagawa et al.,
68 2018), this tool is less commonly used for ecological hypothesis testing, such as the impact
69 of environmental stressors (Li et al., 2018b; Macher et al., 2018).

70 Aquatic ecosystems are highly suited to eDNA studies as eDNA exists in multiple
71 states with rapid modes of transport and degradation, increasing detectability of
72 contemporary biodiversity (Rees et al., 2014; Barnes & Turner, 2015). Lentic systems
73 provide further opportunities for eDNA research, being discrete water bodies with variable
74 physicochemical properties that do not experience flow dynamics (Harper et al., 2019).

75 Ponds in particular have enormous biodiversity and experimental virtue that has not been
76 maximised in previous eDNA metabarcoding assessments of this habitat (Valentini et al.,
77 2016; Evans et al., 2017; Klymus et al., 2017; Ushio et al., 2017; Bálint et al., 2018). These
78 small and abundant water bodies span broad ecological gradients (De Meester et al., 2005)
79 and comprise pondscapes - a network of ponds and their surrounding terrestrial habitat (Hill
80 et al., 2018). Pondscapes contribute substantially to aquatic and non-aquatic biodiversity
81 across spatial scales, with ponds supporting many rare and protected species in fragmented
82 landscapes (De Meester et al., 2005; Biggs et al., 2016; Hill et al., 2018). Consequently,
83 ponds are model systems for experimental validation and examination of biogeographical
84 patterns (De Meester et al., 2005). Habitat complexity and tools required for different taxa
85 with associated bias (Evans et al., 2017) and cost (Valentini et al., 2016) once hindered
86 exhaustive sampling of pond biodiversity (Hill et al., 2018), but eDNA metabarcoding may
87 overcome these barriers (Harper et al., 2019).

88 In the UK, the threatened great crested newt (*Triturus cristatus*) is an umbrella
89 species for pond conservation. The extensive literature on *T. cristatus* ecology provides an
90 excellent opportunity to validate ecological patterns revealed by eDNA metabarcoding. Both
91 biotic (e.g. breeding substrate, prey, and predators) and abiotic (e.g. pond area, depth, and
92 temperature) factors are known to influence *T. cristatus* breeding success (Langton, Beckett
93 & Foster, 2001). The *T. cristatus* Habitat Suitability Index (HSI [Oldham et al., 2000; ARG-UK,
94 2010]) accounts for these factors using 10 suitability indices that are scored and combined
95 to calculate a decimal score between 0 and 1 (where 1 = excellent habitat). Larvae are
96 susceptible to fish and waterfowl predation (Edgar & Bird, 2006; Rannap & Briggs, 2006;
97 Skei et al., 2006; Hartel, Nemes & Oellerer, 2010), and adults reportedly avoid ponds
98 containing three-spined stickleback (*Gasterosteus aculeatus*) (McLee & Scaife, 1992),

99 ninespine stickleback (*Pungitius pungitius*), crucian carp (*Carassius carassius*), and common
100 carp (*Carassius carpio*) (Rannap, Lõhmus & Briggs, 2009a, b). Conversely, *T. cristatus* and
101 smooth newt (*Lissotriton vulgaris*) prefer similar habitat and often co-occur (Rannap &
102 Briggs, 2006; Skei et al., 2006; Rannap et al., 2009a; Denoël et al., 2013; Cayuela et al.,
103 2018). *T. cristatus* individuals thrive in ponds with good water quality as indicated by diverse
104 macroinvertebrate communities (Oldham et al., 2000; Rannap et al., 2009a), and water
105 clarity is important for breeding displays, foraging success, and egg survival (Rannap &
106 Briggs, 2006; Skei et al., 2006). Pond networks encourage *T. cristatus* occupancy (Joly et al.,
107 2001; Rannap et al., 2009a; Hartel et al., 2010; Denoël et al., 2013), but larger pond area
108 discourages presence (Joly et al., 2001). Ponds with heavy shading (Vuorio, Heikkinen &
109 Tikkanen, 2013) or dense macrophyte cover (Rannap & Briggs, 2006; Skei et al., 2006; Hartel
110 et al., 2010) are unlikely to support viable populations. *T. cristatus* individuals also depend
111 on terrestrial habitat, preferring open, semi-rural ponds (Denoël et al., 2013)
112 containing pasture, extensively grazed and rough grassland, scrub, and coniferous and
113 deciduous woodland (Oldham et al., 2000; Rannap & Briggs, 2006; Rannap et al., 2009a;
114 Gustafson, Malmgren & Mikusiński, 2011; Vuorio et al., 2013).

115 We assessed vertebrate communities at the pondscape using a dataset generated by
116 eDNA metabarcoding for over 500 ponds with comprehensive environmental metadata. We
117 validated eDNA metabarcoding as a tool for ecological hypothesis testing, and compared its
118 outputs to previous results generated by established methods. Specifically, we tested biotic
119 (community presence-absence data) and abiotic (environmental metadata on ponds and
120 surrounding terrestrial habitat) determinants of *T. cristatus* at the pondscape - an
121 impractical task by conventional means. Furthermore, we tested the applicability of the HSI
122 to predict eDNA-based *T. cristatus* occupancy. Finally, we assessed the umbrella species

123 status of *T. cristatus* by investigating whether *T. cristatus* presence and the *T. cristatus* HSI
124 score can predict vertebrate species richness of ponds.

125

126

127 **2. Materials and methods**

128

129 **2.1 Samples**

130

131 We repurposed the taxonomically assigned sequence reads from Harper et al. (2018) that
132 were produced using eDNA metabarcoding of pond water to compare qPCR and eDNA
133 metabarcoding for *T. cristatus* detection. Samples from 508 ponds included in Natural
134 England's Great Crested Newt Evidence Enhancement Programme were processed using
135 eDNA metabarcoding alongside 24 privately surveyed ponds. Water samples were collected
136 using established methodology (Biggs et al., 2015), detailed in Supporting Information:
137 Appendix 1. Briefly, 20 x 30 mL water samples were collected from each pond and pooled.
138 Six 15 mL subsamples were taken from the pooled sample and each added to 33.5 mL
139 absolute ethanol and 1.5 mL sodium acetate 3 M (pH 5.2). Subsamples were pooled during
140 DNA extraction to produce one eDNA sample per pond. Targeted qPCR detected *T. cristatus*
141 in 265 (49.81%) ponds (Harper et al., 2018).

142 Environmental metadata (Table S1) were collected for 504 of 532 ponds (Fig. S1) by
143 environmental consultants contracted for Natural England's Great Crested Newt Evidence
144 Enhancement Programme. Metadata included: maximum depth of ponds; pond
145 circumference; pond width; pond length; pond area; pond density (i.e. number of ponds per
146 km²); terrestrial overhang; shading; macrophyte cover; HSI score (Oldham et al., 2000); HSI

147 band (categorical classification of HSI score [ARG-UK, 2010]); pond permanence; water
148 quality; pond substrate; presence of inflow or outflow; presence of pollution; presence of
149 other amphibians, fish and waterfowl; woodland; rough grass; scrub/hedge; ruderals; other
150 terrestrial habitat (i.e. good quality terrestrial habitat that did not conform to
151 aforementioned habitat types); and overall terrestrial habitat quality.

152

153 **2.2 DNA reference database construction**

154

155 A custom, phylogenetically curated reference database of mitochondrial 12S ribosomal RNA
156 (rRNA) sequences for UK fish species was previously constructed for eDNA metabarcoding
157 of lake fish communities (Hänfling et al., 2016). Harper et al. (2018) constructed additional
158 reference databases for UK amphibians, reptiles, birds, and mammals (Supporting
159 Information: Appendix 1). Reference sequences available for species varied across
160 vertebrate groups: amphibians 100.00% ($N = 21$), reptiles 90.00% ($N = 20$), mammals 83.93%
161 ($N = 112$), and birds 55.88% ($N = 621$). Table S2 lists species without database
162 representation, i.e. no records for any species in a genus. Sanger sequences were obtained
163 from tissue of *T. cristatus*, *L. vulgaris*, Alpine newt (*Ichthyosaura alpestris*), common toad
164 (*Bufo bufo*), and common frog (*Rana temporaria*) to supplement the amphibian database
165 (Supporting Information: Appendix 1). The complete reference databases compiled in
166 GenBank format were deposited in a GitHub repository and permanently archived
167 (<https://doi.org/10.5281/zenodo.1188710>) by Harper et al. (2018).

168

169 **2.3 Primer validation**

170

171 Reference databases were combined for *in silico* validation of published 12S rRNA primers
172 12S-V5-F (5'-ACTGGGATTAGATACCCC-3') and 12S-V5-R (5'-TAGAACAGGCTCCTCTAG-3') (Riaz
173 et al., 2011) using ecoPCR software (Ficetola et al., 2010). Set parameters allowed a 50-250
174 bp fragment and three mismatches between each primer and reference sequence. Primers
175 were validated *in vitro* for UK fish by Hänfling et al. (2016) and by Harper et al. (2018) for six
176 UK amphibian species (Fig. S2).

177

178 **2.4 eDNA metabarcoding**

179

180 We used the taxonomically assigned sequence reads generated with vertebrate eDNA
181 metabarcoding by Harper et al. (2018). The eDNA metabarcoding workflow is fully described
182 in Harper et al. (2018) and Supporting Information: Appendix 1. eDNA was first amplified
183 with the aforementioned primers, where PCR positive controls (six per PCR plate; $n = 114$)
184 were cichlid (*Rhamphochromis esox*) DNA (0.284 ng/ μ L) and PCR negative controls (six per
185 PCR plate; $n = 114$) were sterile molecular grade water (Fisher Scientific UK Ltd, UK). PCR
186 products were individually purified using E.Z.N.A[®] Cycle Pure V-Spin Clean-Up Kits (Omega
187 Bio-tek, GA, USA) following the manufacturer's protocol. The second PCR bound Multiplex
188 Identification tags to the purified products. PCR products were individually purified using
189 magnetic bead clean-up and quantified with a Quant-IT[™] PicoGreen[™] dsDNA Assay
190 (Invitrogen, UK). Samples were normalised, pooled, and libraries quantified using a Qubit[™]
191 dsDNA HS Assay (Invitrogen, UK). Libraries were sequenced on an Illumina[®] MiSeq using 2 x
192 300 bp V3 chemistry (Illumina, Inc, CA, USA) and raw sequence reads processed using

193 metaBEAT (metaBarcoding and Environmental Analysis Tool) v0.97.7
194 (<https://github.com/HullUni-bioinformatics/metaBEAT>). After quality filtering, trimming,
195 merging, chimera detection, and clustering, non-redundant query sequences were
196 compared against our reference database using BLAST (Zhang et al., 2000). Putative
197 taxonomic identity was assigned using a lowest common ancestor (LCA) approach based on
198 the top 10% BLAST matches for any query matching with at least 98% identity to a reference
199 sequence across more than 80% of its length. Unassigned sequences were subjected to a
200 separate BLAST against the complete NCBI nucleotide (nt) database at 98% identity to
201 determine the source via LCA as described above. The bioinformatic analysis was archived
202 (<https://doi.org/10.5281/zenodo.1188710>) by Harper et al. (2018) for reproducibility.

203

204 **2.5 Data analysis**

205

206 Analyses were performed in R v.3.4.3 (R Core Team, 2017). Data and R scripts have been
207 deposited in a dedicated GitHub repository for this study, which has been permanently
208 archived at: <https://doi.org/10.5281/zenodo.2634427>. Assignments from different
209 databases were merged, and spurious assignments (i.e. non-UK species, invertebrates and
210 bacteria) removed from the dataset. The family Cichlidae was reassigned to
211 *Rhamphochromis esox*. The green-winged teal (*Anas carolinensis*) was reassigned to *Anas*
212 (Dabbling ducks) because this species is a rare migrant and available reference sequences
213 were identical across our metabarcode to those for mallard (*Anas platyrhynchos*) and
214 Eurasian teal (*Anas crecca*), which are widely distributed across the UK. Scottish wildcat
215 (*Felis silvestris*) does not occur at the sampling localities (Kent, Lincolnshire and Cheshire)
216 and was therefore reassigned to domestic cat (*Felis catus*). Wild boar (*Sus scrofa*) and grey

217 wolf (*Canis lupus*) were reassigned to domestic pig (*Sus scrofa domesticus*) and domestic
218 dog (*Canis lupus familiaris*) given the restricted distribution of *S. scrofa* and absence of *C.*
219 *lupus* in the UK. The genus *Strix* was reassigned to tawny owl (*Strix aluco*) as it is the only UK
220 representative of this genus. Where family and genera assignments containing a single UK
221 representative had reads assigned to species, reads from all assignment levels were merged
222 and manually assigned to that species.

223 Of the 114 PCR negative controls included, 50 produced no reads (Fig. S3). Reads
224 generated for 64 of 114 PCR negative controls ranged from 0 to 49227, and strength of each
225 contaminant varied (mean = 0.021%, range = 0 - 100.0% of the total reads per PCR negative
226 control). To minimise risk of false positives, we evaluated different sequence thresholds.
227 These included the maximum sequence frequency of *R. esox* DNA in eDNA samples (100%),
228 maximum sequence frequency of any DNA except *R. esox* in PCR positive controls (83.96%),
229 and taxon-specific thresholds (maximum sequence frequency of each taxon in PCR positive
230 controls). The different thresholds were applied to the eDNA samples and the results from
231 each compared (Fig. S4). The taxon-specific thresholds (Table S3) retained the most
232 biological information, thus these were selected for downstream analysis. Consequently,
233 taxa were only classed as present at sites if their sequence frequency exceeded their
234 threshold. After applying the taxon-specific thresholds, our PCR positive control (*R. esox*),
235 human (*Homo sapiens*), and domestic species (Table S4) were removed from the dataset.
236 Higher taxonomic assignments excluding the genus *Anas* were then removed, thus
237 taxonomic assignments in the final dataset were predominantly of species resolution.

238 The read count data were converted to a species presence-absence matrix.
239 Hypotheses tested relating to biotic and abiotic determinants of *T. cristatus* occupancy as
240 well as the umbrella status of *T. cristatus* are summarised in Table 1. We employed

241 Generalized Linear Mixed-effects Models (GLMMs) using the package lme4 v1.1-12 (Bates et
242 al., 2015) for hypothesis testing. First, we investigated the influence of vertebrate group
243 species richness on *T. cristatus* occupancy using a binomial GLMM with the logit link
244 function that included species richness of other amphibians, fish, waterfowl, terrestrial
245 birds, and mammals as fixed effects and pond as a random effect ($N = 532$). We performed a
246 preliminary analysis using the package cooccur v1.3 (Griffith, Veech & Marsh, 2016) to
247 identify species associations between *T. cristatus* and other vertebrates ($N = 532$). Identified
248 associations in conjunction with the existing *T. cristatus* literature informed candidate biotic
249 variables to be modelled against *T. cristatus* occupancy ($n = 504$). The existing *T. cristatus*
250 literature informed candidate abiotic variables to be modelled against *T. cristatus*
251 occupancy ($n = 504$).

252 Selection of a suitable set of explanatory variables and modelling framework is fully
253 described in Supporting Information: Appendix 1. Briefly, candidate biotic and abiotic
254 explanatory variables were assessed for collinearity, relative importance, and non-linearity.
255 We constructed separate binomial GLMMs with the logit link function for biotic and abiotic
256 explanatory variables that included sample as a random effect. We modelled HSI score
257 (fixed effect) and sample (random effect) separately to prevent HSI score masking variation
258 caused by the individual biotic and abiotic variables it encompasses. Using a Poisson GLMM
259 with sample as a random effect, we tested the umbrella species status of *T. cristatus* by
260 modelling vertebrate species richness against *T. cristatus* presence-absence and the *T.*
261 *cristatus* HSI score.

262 For each GLMM, we employed an information-theoretic approach using Akaike
263 Information Criterion (AIC) to determine the most parsimonious approximating model to
264 make predictions (Akaike, 1973). Biotic and abiotic models considered respectively were

265 nested thus the best models were chosen using stepwise backward deletion of terms based
266 on Likelihood Ratio Tests (LRTs). The HSI score and vertebrate species richness models were
267 compared to null GLMMs. Final models were tested for overdispersion using a custom
268 function testing overdispersion of the Pearson residuals. Model fit was assessed using the
269 Hosmer and Lemeshow Goodness of Fit Test within the package ResourceSelection v0.2-4
270 (Lele, Keim & Solymos, 2016), quantile-quantile plots, and partial residual plots (Zuur et al.,
271 2009). Model predictions were obtained using the predictSE function in the package
272 AICcmodavg v2.0-3 (Mazerolle, 2016) and upper and lower 95% CIs were calculated from
273 the standard error of the predictions. Results were plotted using the package ggplot2 v2.1.0
274 (Wickham, 2016).

275

276

277 **3. Results**

278

279 **3.1 eDNA metabarcoding**

280

281 A total of 532 eDNA samples and 228 PCR controls were processed across two sequencing
282 runs. The runs generated raw sequence read counts of 36,236,862 and 32,900,914
283 respectively. After quality filtering, trimming, and merging of paired-end reads, 26,294,906
284 and 26,451,564 sequences remained. Following removal of chimeras and redundancy via
285 clustering, the libraries contained 14,141,237 and 14,081,939 sequences (average read
286 counts of 36,826 and 36,671 per sample respectively), of which 13,126,148 and 13,113,143
287 sequences were taxonomically assigned. The final dataset (assignments corrected,
288 thresholds applied, and assignments removed) contained 53 vertebrate species (Table S5),

289 including six amphibians, 14 fish, 17 birds, and 16 mammals (Fig. 1).

290

291 **3.2 Pondscape biodiversity**

292

293 All native amphibians were found as well as the non-native marsh frog (*Pelophylax*
294 *ridibundus*). *T. cristatus* ($n = 148$), *L. vulgaris* ($n = 151$) and *R. temporaria* ($n = 122$) were
295 widespread, but *B. bufo* ($n = 42$), palmate newt (*Lissotriton helveticus* [$n = 5$]) and *P.*
296 *ridibundus* were uncommon ($n = 1$). The threatened European eel (*Anguilla anguilla* [$n =$
297 15]), European bullhead (*Cottus gobio* [$n = 14$]), and *C. carassius* ($n = 2$) were detected
298 alongside native fishes, such as pike (*Esox Lucius* [$n = 17$]) and roach (*Rutilus rutilus* [$n = 71$]),
299 but also introduced species, including *C. carpio* ($n = 40$), ruffe (*Gymnocephalus cernua* [$n =$
300 1]), and rainbow trout (*Oncorhynchus mykiss* [$n = 3$]). Some identified waterfowl were
301 ubiquitous, such as common moorhen (*Gallinula chloropus* [$n = 211$]), whereas others were
302 less common, e.g. grey heron (*Ardea cinerea* [$n = 1$]) and Eurasian oystercatcher
303 (*Haematopus ostralegus* [$n = 1$]). Terrestrial fauna were often detected in fewer than five
304 ponds (Figs. 1c, d). Buzzard (*Buteo buteo* [$n = 4$]), Eurasian jay (*Garrulus glandarius* [$n = 7$]),
305 dunnock (*Prunella modularis* [$n = 4$]), and starling (*Sturnus vulgaris* [$n = 4$]) were the most
306 frequently detected terrestrial birds. Introduced mammals (Mathews et al., 2018), such as
307 grey squirrel (*Sciurus carolinensis* [$n = 57$]) and Reeve's muntjac (*Muntiacus reevesi* [$n = 3$]),
308 outweighed native mammals. Nonetheless, we detected several mammals with Biodiversity
309 Actions Plans and/or of conservation concern (Mathews et al., 2018), including otter (*Lutra*
310 *lutra* [$n = 1$]), water vole (*Arvicola amphibious* [$n = 16$]), European polecat (*Mustela putorius*
311 [$n = 1$]), brown hare (*Lepus europaeus* [$n = 1$]) and water shrew (*Neomys fodiens* [$n = 8$]).
312 Notably, the invasive American mink (*Neovison vison*) was absent despite widespread UK

313 distribution (Mathews et al., 2018). All species and their detection frequencies are listed in
314 Table S5.

315

316 **3.3 Biotic determinants of *T. cristatus* occupancy**

317

318 *T. cristatus* occupancy was positively influenced by amphibian and waterfowl species
319 richness, yet negatively influenced by mammal species richness. *T. cristatus* occupancy was
320 reduced as fish and terrestrial bird species richness increased, but these trends were not
321 significant (Fig. 2; GLMM: overdispersion $\theta = 0.994$, $\chi^2_{525} = 521.778$, $P = 0.532$; fit $\chi^2_8 =$
322 48.537 , $P < 0.001$, $R^2 = 10.91\%$). *T. cristatus* had significant ($P < 0.05$) positive associations
323 with three species (Fig. S5), including *L. vulgaris*, common coot (*Fulica atra*), and *G.*
324 *chloropus*. However, *T. cristatus* had significant ($P < 0.05$) negative associations with six
325 species (Fig. S5), including *B. bufo*, *C. carpio*, *G. aculeatus*, *P. pungitius*, common pheasant
326 (*Phasianus colchicus*), and *S. carolinensis*. Only presence-absence of *L. vulgaris*, *B. bufo*, *C.*
327 *carpio*, *G. aculeatus*, and *G. chloropus* were retained by model selection as explanatory
328 variables for the biotic GLMM of *T. cristatus* occupancy (Figs. 3a-e; GLMM: overdispersion θ
329 $= 1.001$, $\chi^2_{495} = 495.297$, $P = 0.488$; fit $\chi^2_8 = 101.820$, $P < 0.001$, $R^2 = 27.80\%$). Waterfowl
330 presence-absence was also retained in the biotic GLMM (Fig. 3f). Results of analyses are
331 summarised and compared to previously reported determinants in Table 1.

332 *T. cristatus* individuals were more likely to occupy ponds with more amphibian
333 species (Fig. 2a). *T. cristatus* was detected in 51.66% of ponds ($n = 151$) containing *L.*
334 *vulgaris*, but in only 11.91% of ponds ($n = 42$) with *B. bufo* (Fig. 1a). Probability of *T. cristatus*
335 occupancy was lower in ponds with more fish species, and *T. cristatus* was absent from
336 ponds with more than four fish species (Fig. 2b). *T. cristatus* was only found in 15.00% ($n =$

337 40), 14.55% ($n = 55$) and 6.67% ($n = 15$) of ponds inhabited by *C. carpio*, *G. aculeatus* and *P.*
338 *pungitius* respectively (Fig. 1b). In contrast, *T. cristatus* individuals were more likely to occur
339 in ponds with more waterfowl species (Fig. 2c). *T. cristatus* occupied 41.67% ($n = 48$) and
340 36.02% ($n = 211$) of ponds with *F. atra* and *G. chloropus* respectively (Fig. 1c). *T. cristatus*
341 occupancy was negatively influenced by higher terrestrial bird species richness, but not
342 significantly so (Fig. 2d). However, *T. cristatus* was negatively associated with *P. colchicus*,
343 being found in only 12.00% ($n = 25$) ponds with *P. colchicus* records (Fig. 1c). *T. cristatus*
344 were less likely to occupy ponds with more mammal species (Figs. 2e). Specifically, *T.*
345 *cristatus* was negatively associated with *S. carolinensis* and found in only 15.79% ($n = 57$) of
346 ponds with *S. carolinensis* records (Fig. 1d).

347

348 **3.4 Abiotic determinants of *T. cristatus* occupancy**

349

350 Three explanatory variables were retained in the abiotic GLMM explaining *T. cristatus*
351 occupancy (GLMM: overdispersion $\theta = 1.004$, $\chi^2_{499} = 500.995$, $P = 0.467$; fit $\chi^2_8 = 14.409$, $P =$
352 0.072 , $R^2 = 9.73\%$). The probability of *T. cristatus* occupancy decreased in ponds with inflow
353 present, larger area, and higher percentage of shading (Table 1, Figs. 3g-i).

354

355 **3.5 *T. cristatus* HSI and umbrella status**

356

357 HSI score positively correlated with *T. cristatus* occupancy (GLMM: overdispersion $\theta = 1.012$,
358 $\chi^2_{501} = 506.893$, $P = 0.418$; fit $\chi^2_8 = 7.270$, $P = 0.508$, $R^2 = 5.45\%$), where *T. cristatus*
359 individuals were more likely to occupy ponds with a higher HSI score (Table 1, Fig. 3j).
360 Vertebrate species richness was positively associated with *T. cristatus* occupancy (GLMM:

361 overdispersion $\theta = 0.977$, $\chi^2_{500} = 488.687$, $P = 0.633$; fit $\chi^2_8 = -158.03$, $P = 1.000$, $R^2 =$
362 13.63%), with more species detected in ponds occupied by *T. cristatus* (Fig. 4a; $\chi^2_1 = 40.985$,
363 $P < 0.001$). However, vertebrate species richness did not significantly increase with the *T.*
364 *cristatus* HSI score (Fig. 4b; $\chi^2_1 = 1.207$, $P = 0.272$).

365

366

367 **4. Discussion**

368

369 We have validated eDNA metabarcoding for ecological hypothesis testing using the
370 community data generated by this tool in combination with environmental metadata for
371 ponds. We tested biotic and abiotic determinants of *T. cristatus* occupancy, whether the HSI
372 can be applied to *T. cristatus* eDNA survey, and whether *T. cristatus* is truly an umbrella
373 species for pond conservation. *T. cristatus* occupancy was higher in ponds containing *L.*
374 *vulgaris* and *G. chloropus*, and ponds without *B. bufo*, *C. carpio*, and *G. aculeatus*. *T.*
375 *cristatus* individuals were also more likely to occupy ponds where waterfowl occurred.
376 Ponds inhabited by *T. cristatus* were typically small, absent of inflow, and not excessively
377 shaded. The *T. cristatus* HSI was appropriate for predicting *T. cristatus* occupancy, but not
378 vertebrate species richness. Nonetheless, more vertebrates were present in ponds occupied
379 by *T. cristatus* thus presence of this amphibian may indicate good quality habitat for other
380 vertebrates. Our findings demonstrate the power of eDNA metabarcoding to enhance
381 freshwater monitoring and research by providing biodiversity data *en masse* at low cost.

382

383 **4.1 Pondscape biodiversity**

384

385 eDNA metabarcoding detected six amphibian, 14 fish, 17 bird, and 16 mammal species
386 across 532 UK ponds. This diverse species inventory emphasises the importance of ponds as
387 habitat for aquatic taxa, but also as stepping stones for semi-aquatic and terrestrial taxa (De
388 Meester et al., 2005; Hill et al., 2018) through provision of drinking, foraging, dispersive, and
389 reproductive opportunities (Biggs et al., 2016; Klymus et al., 2017). Some species detections
390 may be the result of eDNA transport from water bodies in the surrounding area (Hänfling et
391 al., 2016) to ponds via inflow. However, this signifies the capacity of ponds to provide
392 natural samples of freshwater and terrestrial biodiversity in the wider catchment (Deiner et
393 al., 2017).

394

395 **4.2 Biotic determinants of *T. cristatus* occupancy**

396

397 *T. cristatus* were more likely to occupy ponds with higher amphibian species richness -
398 particularly ponds containing *L. vulgaris* and absent of *B. bufo*. *T. cristatus* and *L. vulgaris*
399 have similar habitat requirements and tend to breed in the same ponds (Skei et al., 2006;
400 Rannap et al., 2009a; Denoël et al., 2013; Cayuela et al., 2018), with >60% overlap reported
401 (Rannap & Briggs, 2006). However, *L. vulgaris* can inhabit a broader range of habitat
402 (Rannap & Briggs, 2006; Skei et al., 2006) than *T. cristatus*, which depends on larger, deeper
403 ponds with abundant macrophytes and no fish located in open, semi-rural landscapes
404 (Denoël et al., 2013). *B. bufo* can inhabit fish-containing ponds (Manenti & Pennati, 2016)
405 and *T. cristatus* may predate *B. bufo* eggs and larvae (Langton et al., 2001). This may explain
406 the negative association between *B. bufo* and *T. cristatus* as opposed to the positively

407 associated *T. cristatus* and *L. vulgaris*.

408 *T. cristatus* occupancy marginally decreased with higher fish species richness, and *T.*
409 *cristatus* was negatively associated with *C. carpio*, *G. aculeatus*, and *P. pungitius*. These
410 fishes are common in and typical of ponds. All *T. cristatus* life stages may be predated by
411 fishes (Langton et al., 2001) and negative effects of fish presence-absence on *T. cristatus*
412 occupancy, distribution, and abundance are repeatedly reported (Joly et al., 2001; Rannap &
413 Briggs, 2006; Skei et al., 2006; Denoël & Ficetola, 2008; Rannap et al., 2009a, b; Hartel et al.,
414 2010; Denoël et al., 2013). *G. aculeatus* predated *T. cristatus* eggs and larvae (McLee &
415 Scaife, 1992; Jarvis, 2010), and has non-consumptive effects on *T. cristatus* embryos (Jarvis,
416 2010). *T. cristatus* larvae were also found to alter their behaviour when exposed to
417 predatory *G. aculeatus* but not non-predatory *C. carassius* (Jarvis, 2012), another fish
418 characteristic of ponds.

419 In our study, we detected *T. cristatus* in 50% of ponds inhabited by *C. carassius*, but
420 <20% of ponds containing large and/or predatory fishes, e.g. *C. carpio*, *G. aculeatus*, *E.*
421 *lucius*. Although fewer ponds contained *C. carassius* than *C. carpio*, *G. aculeatus* or *E. lucius*,
422 previous research also indicates large and/or predatory fish are more detrimental to *T.*
423 *cristatus* occurrence (Skei et al., 2006; Hartel et al., 2010; Chan, 2011). *C. carassius* does not
424 hinder *T. cristatus* oviposition, larval behaviour, or recruitment success (Chan, 2011; Jarvis,
425 2012), or pond invertebrate and macrophyte diversity (Stefanoudis et al., 2017). In contrast,
426 *C. carpio* foraging reduces invertebrate density and macrophyte cover (Maceda-Veiga, López
427 & Green, 2017), which lowers *T. cristatus* reproductive and foraging success and heightens
428 predator exposure (Rannap & Briggs, 2006; Gustafson et al., 2006; Chan, 2011). *C. carassius*
429 and *C. carpio* are both included among fish species assumed to negatively impact *T. cristatus*
430 and whose presence-absence is assessed for the *T. cristatus* HSI (ARG-UK, 2010). However, it

431 is evident that *C. carassius* does not directly predate *T. cristatus* or indirectly alter its
432 behaviour, reproductive success, or habitat. Therefore, we advocate a systematic re-
433 evaluation of problematic fish species for *T. cristatus* conservation.

434 *T. cristatus* was positively associated with waterfowl species richness, namely
435 presence of *F. atra* and *G. chloropus*. These waterfowl species share macrophytes and
436 macroinvertebrates as resources with amphibians, feeding on both directly (Perrow et al.,
437 1997; Paillisson & Marion, 2001; Wallau et al., 2010). *F. atra* and *G. chloropus* crop
438 emergent macrophytes to search for invertebrate prey (Paillisson & Marion, 2001; Wallau et
439 al., 2010), which may indirectly benefit *T. cristatus* foraging. Although *Fulica* spp. can also
440 pull up submerged vegetation and damage vegetation banks (Lauridsen, Jeppesen &
441 Andersen, 1993), diet is macrophyte-dominated in late summer and autumn (Perrow et al.,
442 1997) and unlikely to impact *T. cristatus* breeding in spring (Langton et al., 2001). The
443 positive association identified here between *T. cristatus* and these waterfowl most likely
444 reflects a shared preference for macrophyte-rich ponds.

445 *T. cristatus* were less likely to occupy ponds with higher mammal species richness.
446 Our preliminary cooccur analysis indicated *T. cristatus* had negative associations with *P.*
447 *colchicus* and *S. carolinensis*. The terrestrial associations identified are most likely indirect
448 and a reflection of land-use rather than direct as a result of predation or competition, but
449 further investigation would be worthwhile.

450

451 **4.3 Abiotic determinants of *T. cristatus* occupancy**

452

453 *T. cristatus* was less likely to inhabit large ponds with inflow present and a greater
454 percentage of shading. Although our results indicate *T. cristatus* prefers smaller ponds, pond

455 area does not always influence occupancy (Maletzky, Kyek & Goldschmid, 2007; Denoël &
456 Ficetola, 2008; Gustafson et al., 2011) and was deemed a poor predictor of reproductive
457 success (Vuorio et al., 2013). *T. cristatus* has been found to utilise small and large ponds
458 (Rannap & Briggs, 2006; Skei et al., 2006); however, very small ponds (<124 m²) may be
459 unable to support all life stages, and larger ponds may contain fish and experience
460 eutrophication due to agricultural or polluted run-off (Rannap & Briggs, 2006). Inflow to
461 ponds may exacerbate these problems by facilitating entry of agricultural or polluted run-off
462 and connections to streams and rivers containing large, predatory fish (Freshwater Habitats
463 Trust, 2015). Our results corroborate existing research where viable *T. cristatus* populations
464 were unlikely in ponds that were shaded (Vuorio et al., 2013) or had dense macrophyte
465 cover (Rannap & Briggs, 2006; Skei et al., 2006; Hartel et al., 2010).

466 In our study, most environmental metadata available were qualitative, preventing
467 detailed analyses on pond properties and terrestrial habitat in relation to *T. cristatus*
468 occupancy. Better understanding of *T. cristatus* occupancy in relation to species interactions
469 and habitat quality could be achieved with quantitative data on pond properties (e.g. water
470 chemistry), terrestrial habitat (e.g. type, density, distance to ponds), and aquatic and
471 terrestrial habitat usage by different vertebrate species. Furthermore, given the
472 metapopulation dynamics of *T. cristatus*, future research should investigate spatial drivers
473 (e.g. land cover, pond density, climate variables, roads, rivers, elevation) of *T. cristatus*
474 occupancy using innovative modelling approaches, such as individual-based models
475 (Messenger & Olden, 2018). However, acquiring this data to perform these models is a
476 phenomenal task for large numbers of ponds across a vast landscape (Denoël & Ficetola,
477 2008).

478

479 **4.4 *T. cristatus* HSI and umbrella status**

480

481 We found the HSI can predict eDNA-based *T. cristatus* occupancy at the UK pondscape. This
482 contradicts conventional studies which deemed the index inappropriate for predicting *T.*
483 *cristatus* occupancy or survival probabilities (Unglaub et al., 2015). We detected more
484 vertebrates in ponds containing *T. cristatus*, which may support its status as an umbrella
485 species for pond biodiversity and conservation (Gustafson et al., 2006). We also observed a
486 non-significant increase in vertebrate species richness with increasing *T. cristatus* HSI score.
487 An adapted HSI, designed to predict species richness, could help select areas for
488 management and enhancement of aquatic and terrestrial biodiversity. Until then, presence
489 of *T. cristatus* and its HSI may confer protection to broader biodiversity by identifying
490 optimal habitat for pond creation and restoration to encourage populations of this
491 threatened amphibian. The HSI is not without issue due to qualitative data used for score
492 calculation and subjective estimation of indices (Oldham et al., 2000). For future application
493 of this index in *T. cristatus* eDNA survey, we recommend metabarcoding to quantify some
494 qualitatively assessed indices (e.g. water quality via macroinvertebrate diversity, fish and
495 waterfowl presence) alongside *T. cristatus* detection. Provided rigorous spatial and temporal
496 sampling are undertaken, eDNA metabarcoding can also generate site occupancy data to
497 estimate relative species abundance (Valentini et al., 2016; Hänfling et al., 2016).

498

499 **4.5 Prospects of eDNA metabarcoding for freshwater conservation, management, and** 500 **research**

501

502 We have demonstrated the effectiveness of eDNA metabarcoding for landscape-scale

503 biodiversity monitoring and ecological hypothesis testing. We combined metabarcoding
504 with environmental metadata to revisit hypotheses relating to biotic and abiotic
505 determinants of a threatened amphibian at the UK pondscape. Our findings will guide *T.*
506 *cristatus* conservation in the face of increasing land-use and habitat fragmentation - a
507 poignant issue as protective legislation for this species in the UK is changing. Whilst
508 conservation of threatened species and their habitat should be a priority, the bigger picture
509 should not be ignored. eDNA metabarcoding could enhance our understanding of
510 freshwater networks, particularly pondscales, to enable more effective monitoring,
511 protection, and management of aquatic and terrestrial biodiversity. We are only now
512 beginning to realise and explore these opportunities.

513

514

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516

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523

524

525 **Data Availability**

526

527 The taxonomically assigned sequence reads used in this study were generated by Harper et
528 al. (2018). The raw sequence reads were archived on the NCBI Sequence Read Archive
529 (Bioproject: PRJNA417951; SRA accessions: SRR6285413 - SRR6285678). The bioinformatics
530 analysis was deposited in a GitHub repository and permanently archived
531 (<https://doi.org/10.5281/zenodo.1188710>). R scripts and corresponding data for this study
532 have been deposited in a separate GitHub repository which has been permanently archived
533 (<https://doi.org/10.5281/zenodo.2634427>).

534

535

536 **Author Contributions**

537

538 B.H, L.R.H, L.L.H and N.B conceived and designed the study. H.C.R and N.B contributed
539 samples for processing. L.R.H performed laboratory work and analysed the data. I.P.A and
540 E.L offered advice on and supervised sequencing. C.H assisted with bioinformatics analysis.
541 P.B and S.P contributed datasets for analysis. L.R.H wrote the manuscript, which all authors
542 revised.

543

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769

770 **Table 1.** Summary of biotic and abiotic determinants of *T. cristatus* occupancy. Reported
 771 effects on *T. cristatus* occupancy in the literature and hypothesised effects on eDNA-based
 772 *T. cristatus* occupancy are given for each determinant. Any determinants not reported in the
 773 literature are listed as UNK. Direction of observed effects on eDNA-based *T. cristatus*
 774 occupancy determined by each analysis (GLMM assessing species richness in each
 775 vertebrate group, $N = 532$; preliminary cooccur analysis, $N = 532$; GLMM for biotic factors, n
 776 = 504; GLMM for abiotic factors $n = 504$; and GLMM assessing HSI score, $n = 504$) are given.
 777 No, negative and positive effects are listed as 0, - and + respectively. For categorical
 778 variables with more than one level, effect size and standard error (SE) are only given for
 779 levels reported in the model summary. Test statistic is for LRT used and significant P-values
 780 (<0.05) are in bold. Variables included for model selection but not retained in the final
 781 model are listed as NR. The preliminary cooccur analysis was not applicable (NA) to abiotic
 782 factors.
 783

Determinant	Effect reported	Hypothesised effect	Analysis					
			Cooccur			GLMM		
			Effect	P	DF	Effect size (SE)	χ^2	P
Amphibians								
Species richness	UNK				1	0.514 (0.148)	13.875	<0.001
<i>L. vulgaris</i>	+	+	+	<0.001	1	1.624 (0.302)	50.899	<0.001
<i>B. bufo</i>	UNK		-	0.010	1	-1.799 (0.659)	11.150	0.001
Fish								
Species richness	UNK				1	-0.186 (0.123)	2.415	0.120
<i>G. aculeatus</i>	-	-	-	0.012	1	-1.053 (0.469)	6.302	0.012
<i>C. carpio</i>	-	-	-	0.040	1	-1.463 (0.603)	7.841	0.005
<i>P. pungitius</i>	-	-	-	0.049			NR	
<i>C. carassius</i>	-	-						

Waterfowl

Species richness	UNK				1	0.475 (0.141)	12.715	<0.001
<i>G. chloropus</i>	UNK		+	0.001	1	0.897 (0.251)	15.705	<0.001
<i>F. atra</i>	UNK		+	0.022			NR	
Waterfowl presence	-	-			2		6.352	0.042
Minor						0.534 (0.244)		
Major						0.812 (0.506)		

Terrestrial birds

Species richness	UNK				1	-0.046 (0.302)	0.024	0.878
<i>P. colchicus</i>	UNK		-	0.050				

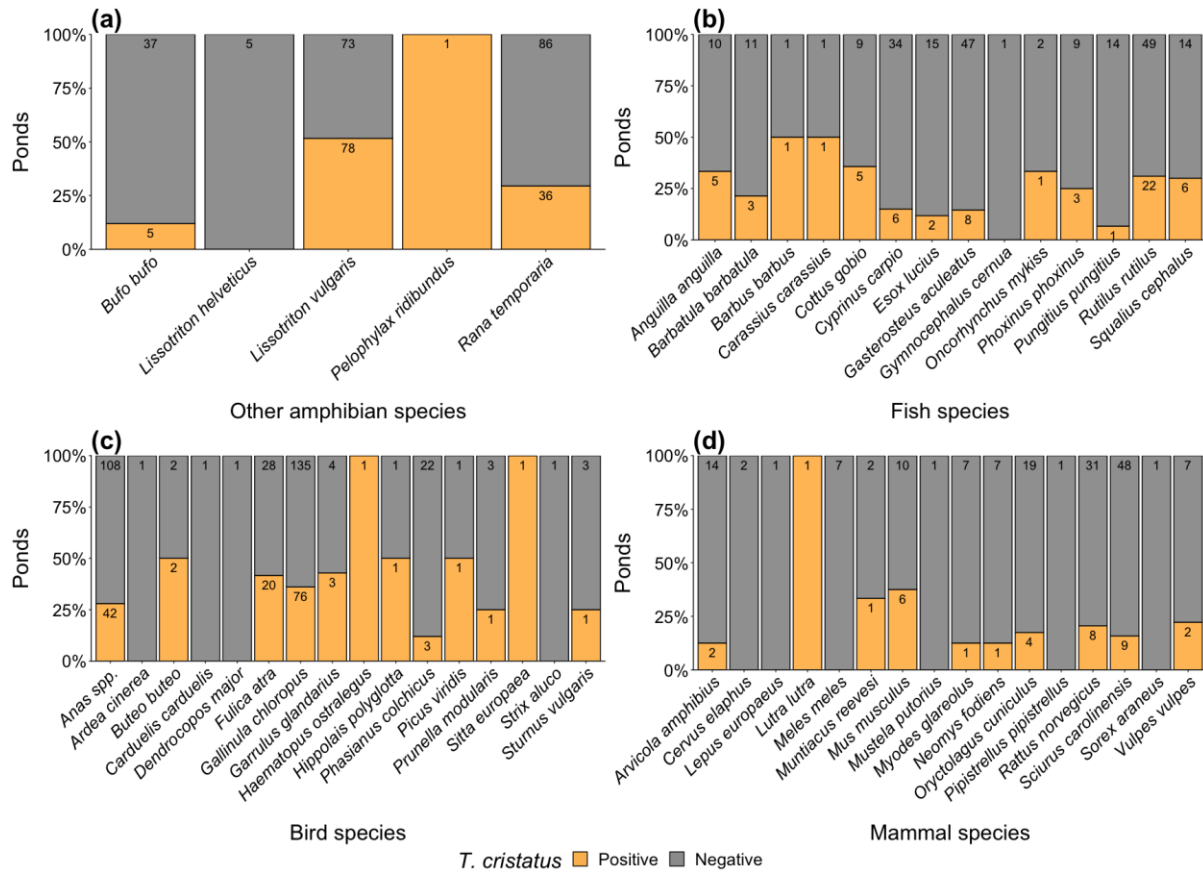
Terrestrial mammals

Species richness	UNK				1	-0.519 (0.193)	8.599	0.007
<i>S. carolinensis</i>	UNK		-	0.020				

Pond area	-/+	-	NA	NA	1	-0.0003 (0.0002)	4.108	0.043
Pond density	+	+	NA	NA			NR	
Pond depth	+	+	NA	NA			NR	
Pond substrate	+	+	NA	NA			NR	
Pond permanence	+	+	NA	NA			NR	
Water quality	+	+	NA	NA			NR	
Inflow Present	-	-	NA	NA	1	-0.887 (0.249)	15.066	<0.001
Outflow	UNK		NA	NA			NR	
Macrophyte cover	-/+	-	NA	NA			NR	
Shading	-/+	-	NA	NA	1	-0.010 (0.003)	12.133	<0.001
Woodland	+	+	NA	NA			NR	

Scrub/hedge	+	+	NA	NA				NR
Ruderals	UNK		NA	NA				NR
HSI score	0/+	+	NA	NA	1	3.164 (0.798)	17.039	<0.001

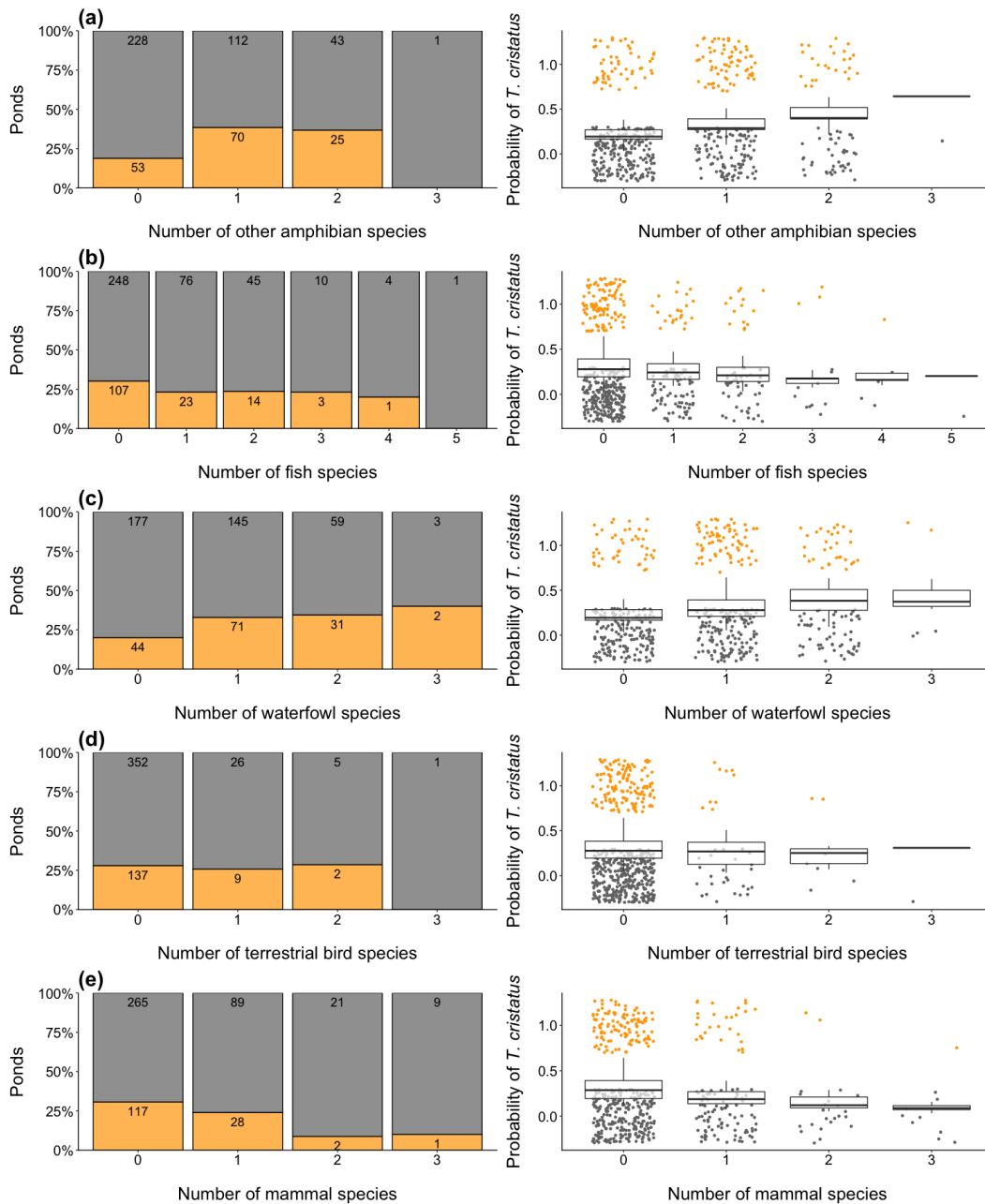
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785

786 **Figure 1.** eDNA metabarcoding detection of *T. cristatus* in relation to other vertebrate
 787 species ($N = 532$ ponds): **(a)** other amphibians, **(b)** fish, **(c)** birds, and **(d)** mammals. Numbers
 788 on each bar are the number of ponds with (orange) and without (grey) *T. cristatus* in which
 789 a vertebrate species was detected.

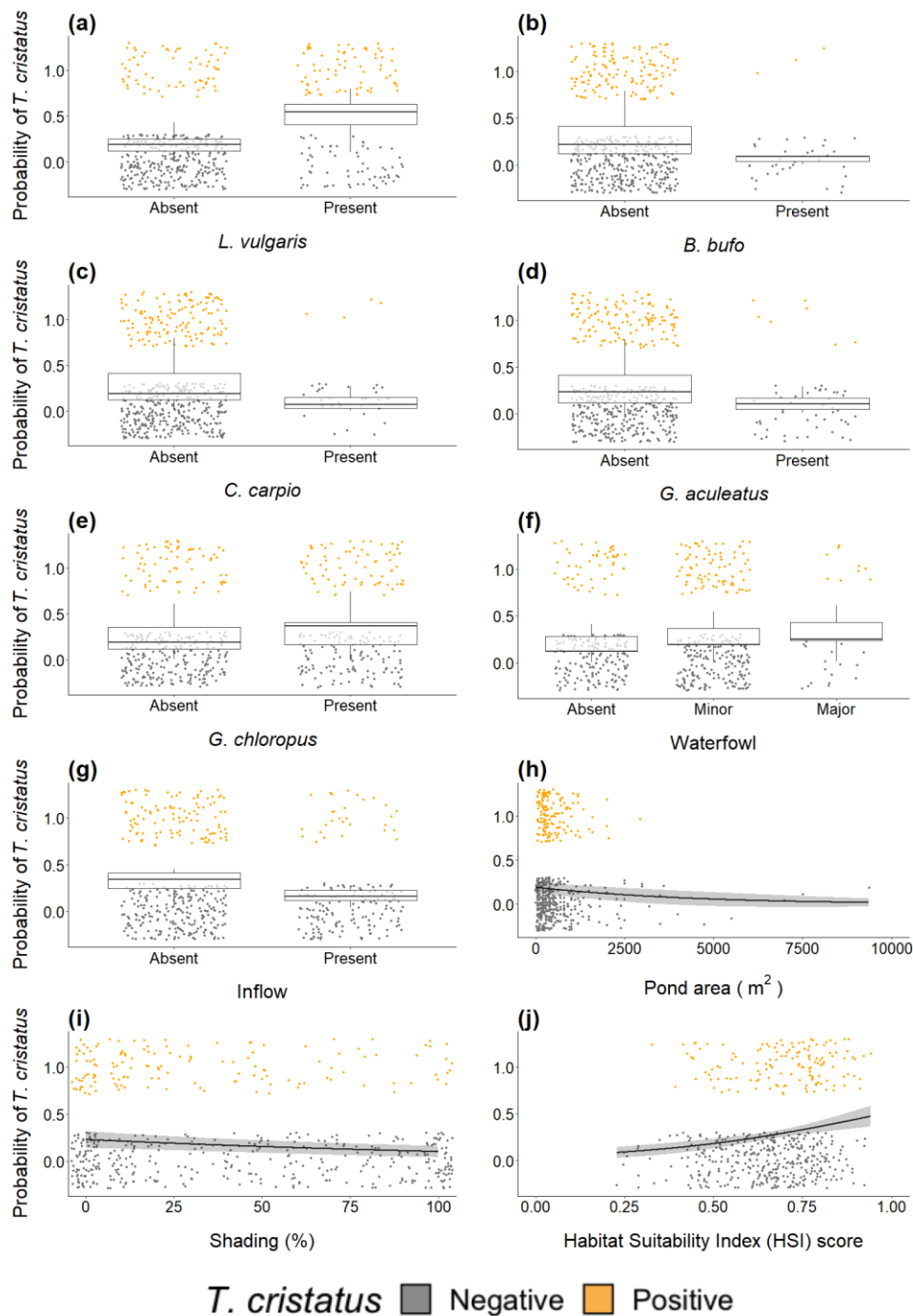
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T. cristatus ■ Negative ■ Positive

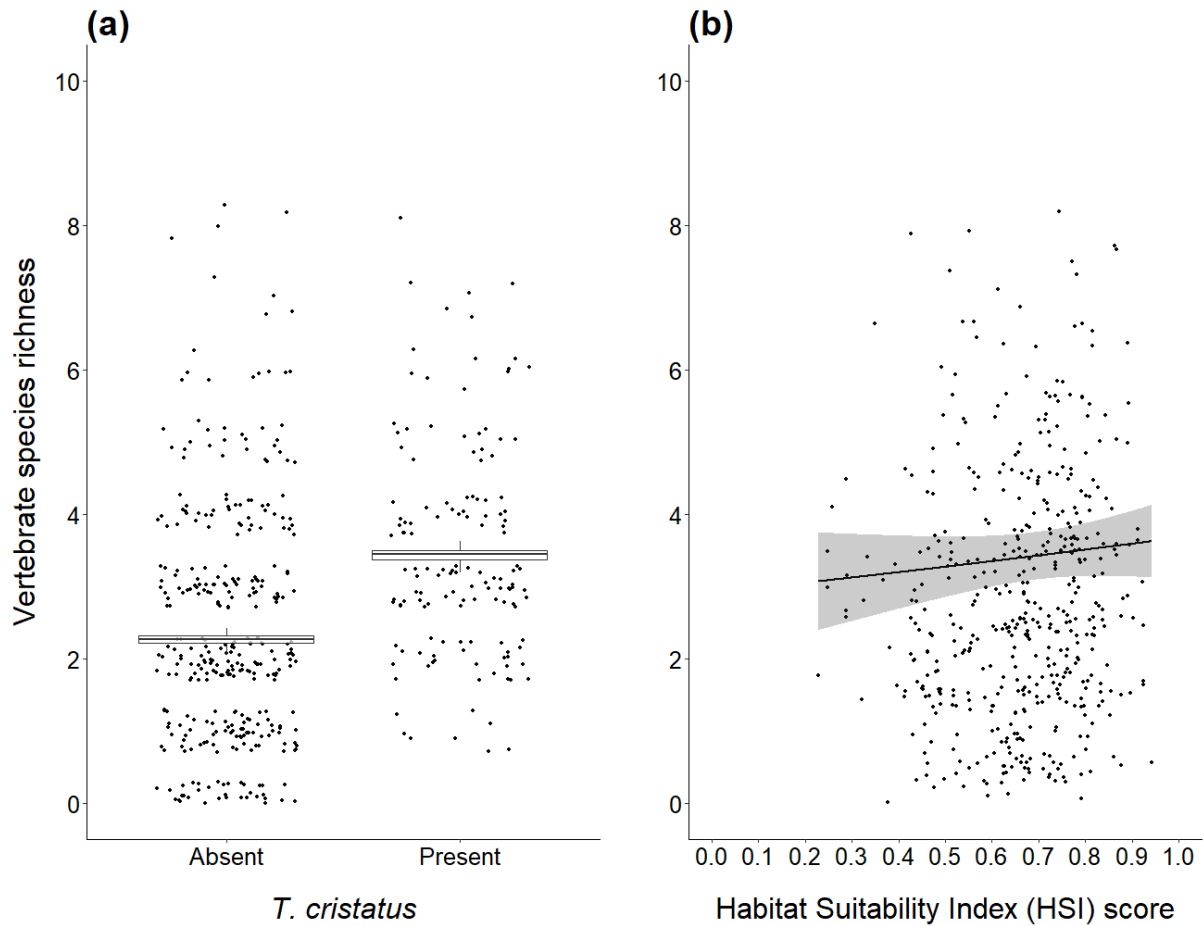
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792 **Figure 2.** *T. cristatus* presence (orange) and absence (grey) in relation to species richness of
 793 different vertebrate groups ($N = 532$ ponds): **(a)** other amphibians, **(b)** fish, **(c)** waterfowl,
 794 **(d)** terrestrial birds, and **(e)** mammals. Observed proportion of ponds with and without *T.*
 795 *cristatus* (left) is plotted alongside predicted probability of *T. cristatus* occupancy (right).
 796 Numbers on barplots of observed occupancy are the number of ponds for each category. In
 797 plots showing predicted *T. cristatus* occupancy, the observed data is shown as points
 798 (jittered around 0 and 1 to clarify variation in point density) and boxes are the model
 799 predictions.



800

801 **Figure 3.** Biotic and abiotic determinants of *T. cristatus* occupancy ($n = 504$ ponds): **(a)** *L.*
 802 *vulgaris* occupancy, **(b)** *B. bufo* occupancy, **(c)** *C. carpio* occupancy, **(d)** *G. aculeatus*
 803 occupancy, **(e)** *G. chloropus* occupancy, **(f)** extent of waterfowl presence, **(g)** presence of
 804 inflow, **(h)** pond area, **(i)** percentage of shading, and **(j)** HSI score. The 95% CIs, as calculated
 805 using the predicted *T. cristatus* probability values and standard error for these predictions,
 806 are given for each relationship. The observed *T. cristatus* presence (orange) and absence
 807 (grey) data are displayed as points (jittered around 0 and 1 to clarify variation in point
 808 density) against the predicted relationships (boxes/lines).



809

810

811 **Figure 4.** Vertebrate species richness in ponds ($n = 504$) in relation to: **(a)** *T. cristatus*
812 occupancy and **(b)** the *T. cristatus* HSI score. The 95% CIs, as calculated using the predicted
813 species richness values and standard error for these predictions, are given for each
814 relationship. The observed data are displayed as points (jittered to clarify variation in point
815 density) against the predicted relationships (boxes/lines).

816