# 1 Diet diversification shapes broad-scale distribution

# 2 patterns in European bats

3 Antton Alberdi<sup>1</sup>, Orly Razgour<sup>2</sup>, Ostaizka Aizpurua<sup>1</sup>, Roberto Novella-Fernandez<sup>2</sup>, Joxerra Aihartza<sup>3</sup>,

4 Ivana Budinski<sup>4</sup>, Inazio Garin<sup>3</sup>, Carlos Ibáñez<sup>5</sup>, Eñaut Izagirre<sup>3,6</sup>, Hugo Rebelo<sup>7</sup>, Danilo Russo<sup>8</sup>, Anton

5 Vlaschenko<sup>9</sup>, Violeta Zhelyazkova<sup>10</sup>, Vida Zrncic<sup>11</sup>, M Thomas P Gilbert<sup>1,12</sup>.

6

#### 7 Abstract

Large-scale species' distributions have been traditionally attributed to physiological
traits related to abiotic factors, while behavioural features linked to biotic interactions
have received little attention. We tested the relationship between trophic and spatial
niche breadths through combining species distribution modelling with dietary DNA
metabarcoding of over 400 bats sampled across Europe belonging to seven species. Our
results point to a causality cascade between hunting plasticity, trophic niche breadth and

14 spatial niche breadth, and thus indicate that behavioral plasticity and dietary

15 diversification can contribute to shaping broad-scale species distributions.

16

#### 17 Affiliations

18 <sup>1</sup>The GLOBE Institute, Faculty of Health and Medical Sciences, University of Copenhagen, Copenhagen, Denmark. 19 <sup>2</sup>University of Southampton, Southampton, United Kingdom, <sup>3</sup>University of the Basque Country, Bilbao, The Basque 20 Country. <sup>4</sup>Department of Genetic Research, Institute for Biological Research "Siniša Stanković", University of 21 Belgrade, Serbia. <sup>5</sup>Doñana Biological Station, CSIC, Seville, Spain. <sup>6</sup>Basque Centre for Climate Change BC3, Leioa, 22 The Basque Country. <sup>7</sup>CIBIO-InBIO, Centro de Investigação em Biodiversidade Recursos e Genéticos, Universidade 23 do Porto, Vairão, Portugal. <sup>8</sup>Wildlife Research Unit, Dipartimento di Agraria, Università degli Studi di Napoli Federico 24 II, Portici (Naples) Italy. 9Bat Rehabilitation Center of Feldman Ecopark, Kharkiv, Ukraine. 10National Museum of 25 Natural History, Bulgarian Academy of Sciences, 1 Tsar Osvoboditel Blvd., 1000 Sofia, Bulgaria. <sup>11</sup>Croatian 26 Biospeleological Society, Zagreb, Croatia. <sup>12</sup>Norwegian University of Science and Technology, University Museum, 27 7491 Trondheim, Norway

# 28 Introduction

29 The characterisation and comparison of trophic niches have been core topics in animal ecology 30 since Hutchinson's conceptualisation of the ecological niche (1). While traditionally considered 31 important only for local-scale species interactions (2), dietary features have also proven relevant 32 for broader scale species distributions (3). Nevertheless, the link between species' trophic 33 niches and large-scale distribution patterns is still inconclusive (4). This could be because diet 34 analyses have traditionally relied on particular diversity metrics that overlook some components 35 of dietary diversity, and because until recently, methodological constraints have limited the 36 possibility of performing broad-scale high-resolution diet studies (5). 37 38 Following the advent of high throughput DNA sequencing-based tools, it is now possible to 39 characterise dietary niches across much larger sample sizes, and at levels of detail never seen 40 before (6). DNA-based diversity assessment also enables comprehensive analysis of trophic 41 variation through considering different components of dietary diversity, such as richness (how 42 many prey are consumed), evenness (the balance of the relative consumption of each prey) and 43 regularity (the degree of similarity across consumed prey) (7,8). Richness, evenness and 44 regularity metrics are positively associated with performance in several ecological systems (9). 45 Hence, we hypothesised that these metrics applied to trophic niches also impact the capacity of 46 animals to thrive in a wider range of environmental conditions, and hence trophic niche breadth 47 could potentially contribute to the shaping of species' distributions.

48

49 To test this, we contrasted broad-scale dietary and spatial niches of a vertebrate system,

50 namely the European bat community. Bats provide an excellent opportunity for understanding

51 dietary and spatial diversity patterns due to their spatial variability and well-studied behavioural

52 traits (10). We collected faecal samples from over 400 bats representing seven species

2

53 captured at 40 locations scattered across the European continent. Faeces of each individual bat 54 were independently analysed through DNA metabarcoding and high throughput sequencing by 55 using two complementary primer sets and three replicates per primer. We used the statistical 56 framework recently developed around Hill numbers (11) to contrast trophic niche measures 57 based on richness (dR), richness+evenness (dRE) and richness+evenness+regularity (dRER). 58 The different species-level trophic niche measures were then statistically related to spatial niche 59 breadth metrics as measured by species distribution modelling, as well as to a range of 60 behavioural traits to assess the causal directionality between dietary diversification and spatial

#### 62 **Results**

niche expansion.

61

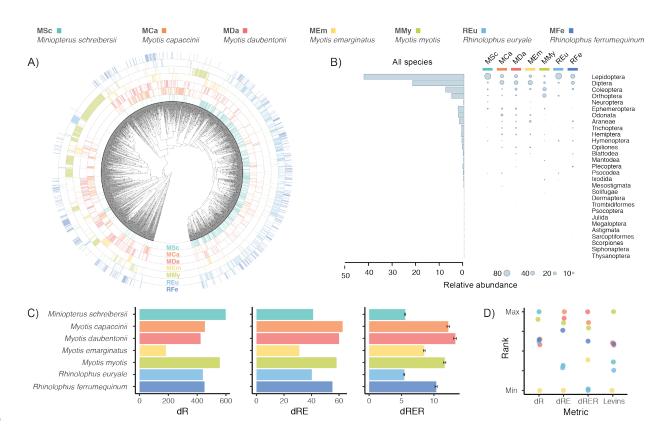
#### 63 The trophic niche of European bats is dominated by Lepidoptera and Diptera

64 After applying all quality filters, the dataset included dietary information of 355 individual bats 65 belonging to seven species (DNA sequencing details in Table S4). Using two primer sets, we 66 detected over 3000 different prev taxa belonging to 29 arthropod orders (Fig. 1A), though the 67 pandiet of European bats was dominated by Lepidoptera and Diptera (Fig. 1B). Our results 68 complement the existing broad-scale molecular dietary data of *Miniopterus schreibersii* (12), 69 and provide the first geographically widespread molecular insights into the dietary ecology of 70 Myotis daubentonii, M. myotis, M. emarginatus, M. capaccinii, Rhinolophus euryale and R. 71 ferrumequinum, which had only been studied at local scales previously (13–15). 72 73 Trophic niche differences depend on the components of diversity accounted for

Trophic niche breadth measures (Fig. 1C), and the species ranks derived from them (Fig. 1D),
were different depending on the components of diversity considered. Similar contrasting results
have also been reported in other systems (16), due to the fact that each diversity component

77 might be driven by different ecological forces (17). For instance, the trophic niches of M. 78 schreibersii and R. euryale showed similar dRE values, yet the contribution of richness and 79 relative evenness components differed. The dietary richness of M. schreibersii was almost 40% 80 larger than R. euryale's, while the evenness factor of R. euryale was almost 30% higher than 81 that of *M. schreibersii*. These differences could be explained by i) the larger home range of *M*. 82 schreibersii compared to R. euryale (18,19), which might expose the former to more prev 83 84 pest moth species in the diet of *M. schreibersii* than in that of *R. euryale* (12,14) —yielding lower 85 relative evenness. As reported for other systems (20), trophic niche differences would be overlooked if the niche breadth analyses were limited to a single diversity metric. 86

87



88

Figure 1. Dietary diversity statistics of the analysed bat species. (A) Radial phylogenetic tree of prey
detected using the Zeale primers and their occurrence patterns in each of the studied bats. A higher

91 resolution image (Fig. S1) and the homologous figure built from the Epp data (Fig. S2) are available in the

92 Supplementary Information. (B) Overall and predator species-specific representation of the arthropod 93 taxonomic orders. (C) Dietary niche breadth measures accounting for richness (dR), richness+evenness 94 (dRE) and richness+evenness+regularity (dRER). The error bars (±SE) of dRER indicate the dispersion 95 of the trophic niche breadths yielded when using different prey phylogenetic trees (N=50) sampled from 96 the Bayesian MCMC. (D) Species ranks according to the dietary niche breadth measure type. Levin's 97 index is also included for being the most common metric employed in the literature.

98

#### 99 Trophic niche breadth explains spatial niche breadth

100 The distribution models generated (Table S5) to test whether trophic niche breadth correlates 101 with spatial niche breadth yielded different spatial projections (Fig. S3) and niche breadth 102 measures for each species. We found that the trophic niche breadth measures accounting for all 103 diversity components (dRER) were positively correlated with the two spatial niche breadth 104 metrics computed, both for each primer-specific dataset (Table S8, Fig. S4) and the overall 105 averaged dataset (Levins' B1: Pearson's r = 0.85; t = 29.57, df = 348, p-value < 0.001; Levins' 106 B2 (Fig. 2A): Pearson's r = 0.80; t = 24.58, df = 348, p-value < 0.001). The species that 107 consume a wider variety of distinct prey are the ones that exhibit broader spatial niches.

108

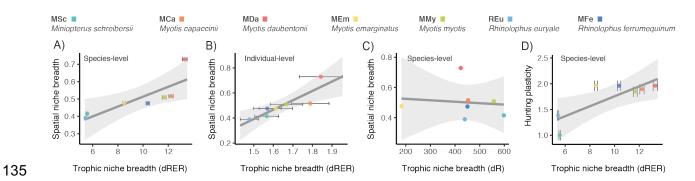
109 We performed a range of analyses to further assess the plausibility of the causal relationship 110 between dietary diversification and spatial niche breadth. We first assessed whether dietary 111 breadths could be broadened passively as a result of spatial niche expansion (21). If the ability 112 to thrive in more distinct environments was driving dietary expansion, we would expect species 113 dietary breadth to be driven by beta diversity, i.e. dietary differences across individuals within 114 species. However, we observed that the correlation between trophic and spatial niche breadth 115 remained significant (Pearson's r = 0.20, t = 3.78, df = 353, p-value < 0.001; Fig. 2B), which 116 highlights the relevance of alpha (individual) dietary diversity. Additionally, if dietary breadths 117 were passively broadened due to spatial niche expansion, we would expect species dietary

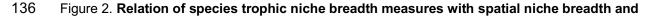
118richness to also increase, as predators are exposed to a larger variety of potential prey.119Nevertheless, we observed that the correlation between trophic and spatial niche breadths120disappeared when relative evenness and regularity components were removed (Levins' B1:121Pearson's r = -0.09; t = -0.20, df = 5, p-value = 0.846; Fig. 2C). Consequently, these two122observations rule out the possibility that dietary breadths are passively broadened as a result of123spatial niche expansion.

124

125 We then investigated the relation between different behavioural traits (hunting plasticity, habitat 126 use diversity and roosting plasticity) and spatial niche breadths to assess whether a third factor 127 could be shaping both dietary and spatial niches. We found no significant correlation between 128 any of the analysed traits and spatial niche breadth (Table S9), yet we found that hunting 129 plasticity is positively related with the trophic niche breadth of predator species (Pearson's r = 130 0.79, t = 24.347, df = 348, p-value < 0.001; Fig. 2D). This suggests that the ability to use a more 131 diverse range of hunting strategies, such as capturing prey from the ground, foliage or water 132 surface, in addition to hunting flying prey (10), broadens the functional spectrum of captured 133 prey, which in turn widens the spatial niche.







137 hunting plasticity. (A) Species-level relation between niche breadth measures accounting for the three

138 components of diversity (dRER: richness+evenness+regularity) and spatial niche breadths of the

139 respective species. (B) Individual-level relation between dRER niche breadth measures and spatial niche

140 breadths of the respective species. (C) Species-level relation between niche breadth measures 141 accounting for only one component of diversity (dR: richness) and spatial niche breadths of the respective 142 species. (D) Species-level relation between dRER niche breadth measures and hunting plasticity of the 143 respective species. Dots indicate mean values per bat species. Note that error bars (±SE) at the species-144 level charts indicate the dispersion of the different trophic niche breadth values yielded from the 50 145 iterations run with different prey phylogenetic trees to account for phylogenetic uncertainty. In contrast, 146 the error bars at the individual-level chart indicates the dispersion of the individual bats' trophic niche 147 breadth values. Chart C does not contain error bars as for dR a single niche breadth measure was 148 computed for each species.

### 149 Discussion

150 This is the first time that high-resolution trophic (DNA metabarcoding) and spatial (Species 151 Distribution Modelling) niche characterisation identify diet as a driving factor of broad-scale 152 spatial patterns. While ecological niche breadth has been previously shown to predict 153 geographical range sizes, trophic niche breadth has so far only been associated with broad-154 scale spatial patterns in arthropods (4). The causality cascade between hunting plasticity. 155 dietary niche breadth and spatial niche breadth seems plausible and is ecologically meaningful. 156 Hunting plasticity has previously been linked to adaptability (22), and could directly affect the 157 fitness of bats, for instance by enabling shifting diets when specific prey types become scarce. 158 Trophic niche breadth could also indirectly affect the fitness of bats by, for example, fostering 159 gut microbiome diversification and dynamism, which have been associated with adaptation 160 capacity in vertebrates (23).

161

Our results contradict the Eltonian noise hypothesis, which proposes that biotic interactions do
not affect species distributions at large geographical scales (2). However, in this case spatial
patterns would not primarily depend on the availability of resources as previously shown (3), but

165 on the inherent behavioural properties of predators. It is noteworthy that all the patterns we 166 found in this study were recovered from trophic niche measures that considered all three 167 components of diversity. No diversity component alone was correlated with either hunting 168 plasticity or spatial niche breadth, which highlights the importance of accounting for relative 169 evenness and regularity of prey when measuring trophic niches. It is also remarkable that a 170 single snapshot of the diet of individual bats was enough to recover a clear link between trophic 171 and spatial niche breadth, as niche patterns are not always coupled at individual and population 172 levels (24). Overall, our study demonstrates the potential of combining environmental DNA with 173 species distribution modelling and behavioural ecology, to unveil broad-scale ecological 174 patterns and links between different components of the ecological niche of species. Finally, our 175 results also highlight the relevance of diet in shaping broad-scale animal distributions, which 176 supports the use of behavioural plasticity as a relevant feature to predict species' range shifts in 177 response to climate change (25).

### 178 Methods

#### 179 Data collection and generation

180 We collected droppings from 402 individual bats captured in 40 locations distributed across 181 Europe (Table S1), in June-October of 2015-2017. The droppings belonged to seven species: 182 Miniopterus schreibersii (MSc), Myotis capaccinii (MCa), Myotis daubentonii (MDa), Myotis 183 emarginatus (MEm), Myotis myotis (MMy), Rhinolophus euryale (REu) and Rhinolophus 184 ferrumequinum (RFe). Using a randomised setup, DNA was extracted from all individual 185 samples and amplified in three replicates using two primer pairs, referred to as Zeale (26) and 186 Epp (27). Amplicons were purified, pooled and built into libraries before Illumina MiSeg 187 sequencing. To ensure maximum DNA sequence reliability, only high quality sequences that 188 appeared in at least two of the three PCR replicates were retained, and sequences identical to

189 those detected in the extraction and library blanks of the corresponding processing batch of 190 each sample were removed. Appropriate sampling depth per sample was ensured by discarding 191 samples with insufficient sequencing depth as assessed by rarefaction curves and curvature 192 indexes. DNA sequences were clustered into operational taxonomic units (OTUs) based on 193 98% identity following Alberdi et al. (28) and taxonomy was assigned by aligning the OTU 194 representative sequences to the Genbank nt (29) — and in the case of Zeale also BOLD (30)— 195 databases. Full details of the field, laboratory and bioinformatics methodologies are reported in 196 the Supplementary Information and Supplementary Code 1.

197

#### 198 Data analysis

199 Diversity analyses were carried out using the R package hilldiv (31) based on abundance-based 200 Hill numbers (32,33). The Hill numbers framework enables i) the relative weight given to 201 abundant and rare OTUs to be modulated through a single parameter, namely the order of 202 diversity q (32), and ii) the similarity level across OTUs to be overlooked or accounted for when 203 computing diversity. Although functional diversities can be computed using Hill numbers (34), 204 given the infeasibility of gathering ecological trait information of thousands of previtems, OTU phylogenies were employed as proxies of ecological resemblance across OTU. Hence, dR 205 206 (richness) was computed as the neutral Hill number of order of diversity q=0; dRE 207 (richness+evenness) was computed as the neutral Hill number of order of diversity q=1 —i.e. 208 Shannon diversity— and dRER (richness+evenness+regularity) was computed as the 209 phylogenetic Hill number of order of diversity g=1. Phylogenetic Hill numbers were computed 210 based on Bayesian phylogenies generated from metabarcoding DNA sequences, and the 211 analyses accounted for the phylogenetic uncertainty of generated trees, as detailed in 212 Supplementary Information. Sample size appropriateness was assessed by comparing 213 observed vs. estimated trophic niche breadth values (Table S14).

214

9

215	Predator species' hunting strategy, habitat-use and roosting data were gathered from 45 articles
216	available in the literature (Tables S12-14). Plasticity indices were computed by means of
217	Shannon diversity of ecological traits. The species distributions models that characterised the
218	spatial niche of predator species were generated using BIOMOD (35), and niche breadths were
219	measured by means of Levins' B1 and B2 metrics (36) based on the spatial projections using
220	the R version ENMTools (37). For all statistical tests, significance threshold was set at p=0.05.
221	All statistical analyses were performed in R (38) after averaging the results yielded by both
222	primers unless otherwise stated.

## 223 Data availability

The datasets generated during and/or analysed during the current study are available in the Dryad repository (ref. [to be included in the reference list when a DOI is available]).

### 226 Code availability

227 The bash, python and R scripts used for analysing the data during the current study are

available in the Supplementary Files as Supplementary Code 1 (DNA metabarcoding),

229 Supplementary Code 2 (Species Distribution Modelling) and Supplementary Code 3 (ecological

230 niche statistical analyses).

## 231 Author contributions

A.A., O.A. and M.T.P.G designed the study. All authors participated in the sample and data
collection. A.A. and O.A. performed the laboratory procedures. O.R. and R.N.F. carried out the
species distribution modelling. O.A. performed the ecological trait analyses. A.A. performed the

- 235 DNA metabarcoding and statistical analyses. A.A. wrote the manuscript. All authors contributed
- to and approved the final version.

### 237 Competing interests

238 The authors declare no competing interests.

### 239 Acknowledgements

- A.A. was supported by Lundbeckfonden (R250-2017-1351) and the Danish Council for
- 241 Independent Research (DFF 5051-00033). O.R. was supported by a NERC Independent
- 242 Research Fellowship (NE/M018660/1), and O.A. was supported by the Carlsberg Foundation's
- 243 Postdoctoral Fellowship (CF15-0619). M.T.P.G. acknowledges ERC Consolidator Grant
- 244 (681396-Extinction Genomics). We are grateful to Fiona Mathews, Daniel Whitby, Roger
- 245 Ransome, Matt Cook and Martina Spada for providing samples; and Aitor Arrizabalaga, Lide
- 246 Jimenez, Vilalii Hukov, Olena Holovchenko, Vanessa Mata, and Branka Pejić for assistance in

the field work.

### 248 References

- Hutchinson, G. E. Concluding remarks. *Cold Spring Harbor Symposium on Quantitative Biology* 22, 415–427 (1957).
- Soberón, J. & Nakamura, M. Niches and distributional areas: concepts, methods, and
   assumptions. *Proc. Natl. Acad. Sci. U. S. A.* **106**, 19644–19650 (2009).
- 253 3. de Araujo, C. B., Marcondes-Machado, L. O. & Costa, G. C. The importance of biotic
- 254 interactions in species distribution models: a test of the Eltonian noise hypothesis using
- parrots. Journal of Biogeography **41**, 513–523 (2014).

- Slatyer, R. A., Hirst, M. & Sexton, J. P. Niche breadth predicts geographical range size: a
   general ecological pattern. *Ecol. Lett.* 16, 1104–1114 (2013).
- 258 5. Alberdi, A. *et al.* Promises and pitfalls of using high-throughput sequencing for diet analysis.
- 259 *Molecular Ecology Resources* **19**, 327–348 (2019).
- 260 6. Bohmann, K. *et al.* Environmental DNA for wildlife biology and biodiversity monitoring.
- 261 *Trends in Ecology Evolution* **29**, 358–367 (2014).
- 262 7. Tucker, C. M. et al. A guide to phylogenetic metrics for conservation, community ecology
- and macroecology. *Biological reviews of the Cambridge Philosophical Society* **92**, 698–715
- 264 (2017).
- 8. Alberdi, A. & Gilbert, M. T. P. A guide to the application of Hill numbers to DNA based
  diversity analyses. *Molecular Ecology Resources* (2019).
- 267 9. Zhang, Y., Chen, H. Y. H. & Reich, P. B. Forest productivity increases with evenness,
- species richness and trait variation: a global meta-analysis. *Journal of Ecology* **100**, 742–
  749 (2012).
- 270 10. Schnitzler, H.-U., Moss, C. F. & Denzinger, A. From spatial orientation to food acquisition in
  271 echolocating bats. *Trends in Ecology and Evolution* **18**, (2003).
- 272 11. Chao, A., Chiu, C.-H. & Jost, L. Unifying species diversity, phylogenetic diversity, functional
- diversity, and related similarity and differentiation measures through Hill Numbers. *Annu.*
- 274 Rev. Ecol. Evol. Syst. 45, 297–324 (2014).
- 12. Aizpurua, O. *et al.* Agriculture shapes the trophic niche of a bat preying on multiple pest
- arthropods across Europe: evidence from DNA metabarcoding. *Molecular Ecology* 27, 815–
  825 (2018).
- 278 13. Vesterinen, E. J. *et al.* What you need is what you eat? Prey selection by the bat Myotis
  279 daubentonii. *Mol. Ecol.* 25, 1581–1594 (2016).
- 280 14. Arrizabalaga-Escudero, A. *et al.* Assessing niche partitioning of co-occurring sibling bat
- 281 species by DNA metabarcoding. *Mol. Ecol.* 27, 1273–1283 (2018).

- 282 15. Galan, M. *et al.* Metabarcoding for the parallel identification of several hundred predators
- and their prey: Application to bat species diet analysis. *Mol. Ecol. Resour.* 18, 474–489
  (2018).
- 16. Ma, M. Species richness vs evenness: independent relationship and different responses to
  edaphic factors. *Oikos* 111, 192–198 (2005).
- 17. Wilsey, B. & Stirling, G. Species richness and evenness respond in a different manner to
  propagule density in developing prairie microcosm communities. *Plant Ecol.* **190**, 259–273
  (2007).
- 290 18. Russo, D., Jones, G. & Migliozzi, A. Habitat selection by the Mediterranean horseshoe bat,
- 291 Rhinolophus euryale (Chiroptera: Rhinolophidae) in a rural area of southern Italy and
- implications for conservation. *Biol. Conserv.* **107**, 71–81 (2002).
- 293 19. Vincent, S., Nemoz, M. & Aulagnier, S. Activity and foraging habitats of Miniopterus
  294 schreibersii (Chiroptera: Miniopteridae) in southern France: implications for its
  295 conservation. *Hystrix* 22, (2010).
- 296 20. MacDonald, Z. G., Nielsen, S. E. & Acorn, J. H. Negative relationships between species
- richness and evenness render common diversity indices inadequate for assessing longterm trends in butterfly diversity. *Biodivers. Conserv.* 26, 617–629 (2017).
- 299 21. Jahner, J. P., Bonilla, M. M., Badik, K. J., Shapiro, A. M. & Forister, M. L. Use of exotic
  300 hosts by Lepidoptera: widespread species colonize more novel hosts. *Evolution* 65, 2719–
  301 2724 (2011).
- 302 22. Manenti, R., Denoël, M. & Ficetola, G. F. Foraging plasticity favours adaptation to new
  303 habitats in fire salamanders. *Anim. Behav.* 86, 375–382 (2013).
- 304 23. Alberdi, A., Aizpurua, O., Bohmann, K., Zepeda-Mendoza, M. L. & Gilbert, M. T. P. Do
  305 vertebrate gut metagenomes confer rapid ecological adaptation? *Trends in Ecology and*
- 306 *Evolution* **31**, 689–699 (2016).
- 307 24. Bolnick, D. I. et al. Ecological release from interspecific competition leads to decoupled

- 308 changes in population and individual niche width. *Proc. Biol. Sci.* 277, 1789–1797 (2010).
- 309 25. Beever, E. A. *et al.* Behavioral flexibility as a mechanism for coping with climate change.
- 310 Front. Ecol. Environ. **15**, 299–308 (2017).
- 311 26. Zeale, M., Butlin, R. K. & Barker, G. Taxon-specific PCR for DNA barcoding arthropod prey
- 312 in bat faeces. *Mol. Ecol.* **11**, 236-244 (2011).
- 313 27. Epp, L. S. et al. New environmental metabarcodes for analysing soil DNA: potential for
- 314 studying past and present ecosystems. *Mol. Ecol.* **21**, 1821–1833 (2012).
- 315 28. Alberdi, A., Aizpurua, O. & Gilbert, M. T. P. Scrutinizing key steps for reliable
- 316 metabarcoding of environmental samples. *Methods Ecol. Evol.* **9**, 134-147 (2018).
- 317 29. Benson, D. A. et al. GenBank. Nucleic Acids Res. 41, D36–42 (2013).
- 318 30. Ratnasingham, S. & Hebert, P. D. N. BOLD: The Barcode of Life Data System
- 319 (www.barcodinglife.org). *Mol. Ecol. Notes* **7**, 355–364 (2007).
- 320 31. Alberdi, A. hilldiv: an R package for integral analysis of diversity based on Hill numbers.
- 321 *bioRxiv* (2019). doi:10.1101/545665
- 322 32. Jost, L. Entropy and diversity. *Oikos* **113**, 363–375 (2006).
- 323 33. Chao, A., Chiu, C.-H. & Jost, L. Phylogenetic diversity measures based on Hill numbers.
  324 *Philos. Trans. R. Soc. Lond. B Biol. Sci.* 365, 3599–3609 (2010).
- 325 34. Chao, A. *et al.* An attribute-diversity approach to functional diversity, functional beta
  326 diversity, and related (dis)similarity measures. *Ecol. Monogr.* **41**, 40 (2019).
- 327 35. Thuiller, W., Lafourcade, B., Engler, R. & Araújo, M. B. BIOMOD a platform for ensemble
   328 forecasting of species distributions. *Ecography* **32**, 369–373 (2009).
- 329 36. Levins, R. *Evolution in Changing Environments: Some Theoretical Explorations*. (Princeton
  330 University Press, 1968).
- 331 37. Warren, D. L., Glor, R. E. & Turelli, M. ENMTools: a toolbox for comparative studies of
  and a studies of environmental niche models. *Ecography* 35, 1160 (2010).
- 333 38. R Development Core Team. *R: a language and environment for statistical computing*.

334 (Vienna, Austria, 2008). at <http://www.R-project.org>