

# Using bats as surrogates to effectively target global hotspots for subterranean conservation and monitoring

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## Research Article

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1 **Using bats as surrogates to effectively target global hotspots for**  
2 **subterranean conservation and monitoring**

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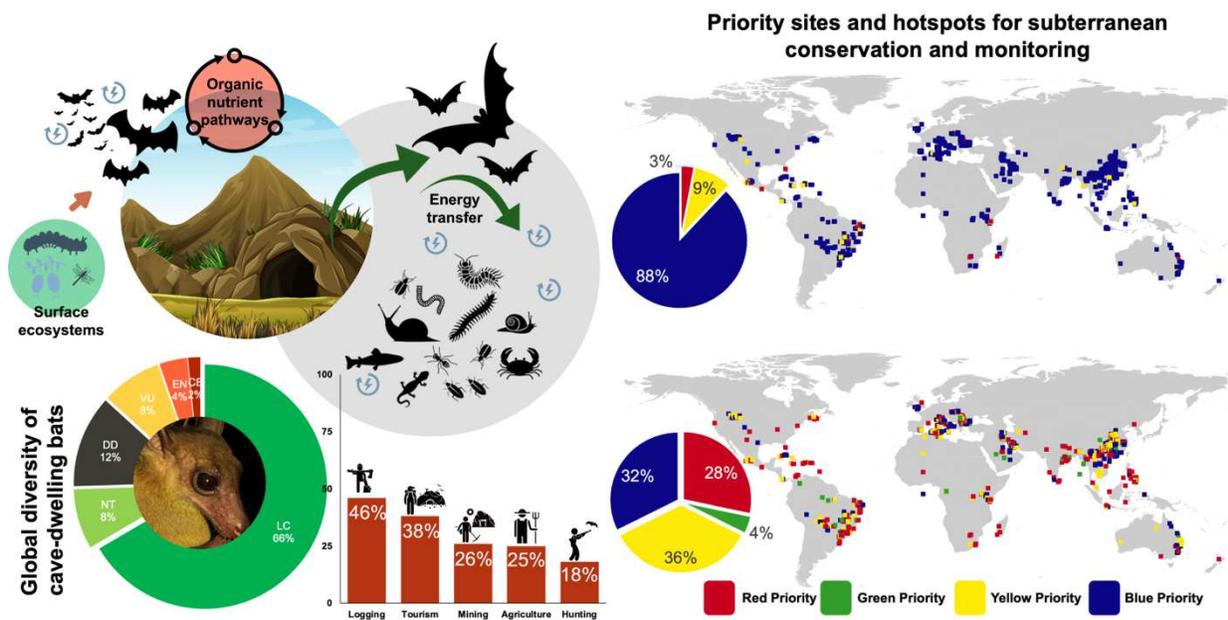
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# 28 Using bats as surrogates to effectively target global hotspots for 29 subterranean conservation and monitoring

## 30 31 Highlights

- 32 • As the second-largest mammal taxa bats provide important ecosystem services  
33 and are keystone to cave ecosystems.
- 34 • Almost half of the global bats are dependent on caves and subterranean  
35 habitats.
- 36 • We include 679 bats species and 1930 caves from 46 countries to assay patterns  
37 of vulnerability and diversity of subterranean habitats across the world.
- 38 • Up to 28% of caves are of high priority.
- 39 • Of high priority caves 87.5% are threatened, and particularly in the tropics (15%).

## 40 41 42 Graphical Abstract



44

45

46 **Abstract**

47 Research and media attention is disproportionately focused on taxa and  
48 ecosystems perceived as charismatic, while other equally diverse systems such as caves  
49 and subterranean ecosystems are often neglected in biodiversity assessments and  
50 prioritisations. Yet these more challenging systems are also vulnerable, with karsts for  
51 example losing around 6% of their area each year, highlighting the urgent need for  
52 protection, especially as up to 90% of cave endemic species may be undescribed. Bats  
53 are keystone to cave ecosystems making them potential surrogates to understand cave  
54 diversity patterns and identify conservation priorities. On a global scale, almost half (48%)  
55 known bat species use caves for parts of their life histories, with 32% endemic to a single  
56 country, and 15% currently threatened. We combined global analysis of cave bats from  
57 the IUCN spatial data with site specific analysis of 1930 bat caves from 46 countries to  
58 develop global priorities for the conservation of the most vulnerable cave ecosystems.  
59 Globally, 28% of caves showed high diversity and were highly threatened and 4% had  
60 high diversity but not currently threatened. Amongst regions, the highest concentration of  
61 conservation priority caves was in the Palearctic and tropical regions (except the  
62 Afrotropics, which requires more intensive data sampling). Our results further highlight  
63 the importance of prioritising bat caves by incorporating locally collected data, and  
64 optimising parameter selection (i.e., appropriate landscape features and threats). Finally,  
65 to protect and conserve these ecosystems it is crucial that we use frameworks such as  
66 this to identify priorities in species and habitat-level, and map vulnerable habitats with the  
67 highest biodiversity and distinctiveness.

68 **Keywords:** Bioindicators, Evolutionary Distinctiveness, Extinction risks, Island  
69 endemism, Subterranean habitats.

## 70 **Introduction**

71 We are currently facing the sixth mass extinction, with a higher rate of extinction  
72 than at any time since former mass extinctions millions of years ago (Vos et al., 2015).  
73 Global biodiversity is threatened by a variety of drivers, and understanding these is critical  
74 to determining effective conservation measures in different systems and scales (Ripple  
75 et al., 2017). Appropriate evidence-based strategies are essential to optimise the  
76 implementation and effectiveness of conservation (Ceballos et al., 2015; Pimm et al.,  
77 2014). Various methods and frameworks have been developed to prioritise taxa and their  
78 habitats for conservation such as habitat prioritisation and zoning (Conenna et al., 2017;  
79 Hernández-Quiroz et al., 2018; Wintle et al., 2019), but many countries may have limited  
80 access to such data or resources to effectively implement conservation (Chandra and  
81 Idrisova, 2011). Priorities should be matched with future targets and effort allocation to  
82 support reaching such targets and ensure that species and their ecosystems are well  
83 protected (Arponen, 2012; Groves et al., 2002; Hughes, 2017; Sutherland et al., 2019).  
84 However, many species lack the data needed for effective priorities to be developed, and  
85 even fewer receive adequate levels of attention (Dolman et al., 2012; Halpern et al.,  
86 2006). Yet, as much as conservation scientists try to address the most pressing  
87 challenges in global biodiversity conservation, many taxa and their associated habitats  
88 are overlooked (Clark and May, 2002). Whilst approaches such as the IUCN Red List of  
89 ecosystems, and the new ecosystem typology goes some-way enabling proactive  
90 targeting of ecosystems, including those with high numbers of endemic species, they may

91 overlook certain types of habitats, particularly those which require significant expertise  
92 and may be challenging to collate data from (Keith et al., 2015).

93         Human activities have already transformed at least 70% of terrestrial ecosystems  
94 to support human populations (Ellis et al., 2013). Modern combined approaches and  
95 technologies are applied to understand environmental processes and threats on above-  
96 ground surface ecosystems (e.g., forest ecosystems) which can easily be mapped with  
97 remote sensing (Rose et al., 2015). Yet most subterranean habitats including caves are  
98 challenging to map, and have frequently been overlooked and neglected in prioritisation  
99 (Mammola et al., 2019a; McClure et al., 2020; Sánchez-Fernández et al., 2021). This  
100 becomes a barrier to research leaving other habitats and the species dependent on them  
101 such as bats understudied and under-protected (Cardinale et al., 2012). The lack of  
102 research necessary to understand the unique adaptations and interactions in these  
103 systems further challenged subterranean research, which not only prevents cave taxa  
104 from garnering public attention, but the special skills needed to study many cave taxa  
105 further add to the lack of existing efforts. Cave ecosystems host a variety of highly  
106 adapted and sensitive organisms, many of which are cave obligates and dependent on  
107 bat guano for nutrients (Deharveng and Bedos, 2018; Ferreira, 2019; Furey and Racey,  
108 2016; Monro et al., 2018a; Simon, 2019), and these studies show that guano is the main  
109 source of nutrition in most of these systems. Some cave systems have relatively high  
110 endemism and up to 90% of species in a single cave may be undescribed (Monro et al.,  
111 2018b; Whitten, 2009), meaning many thousands of species remain undescribed and  
112 potentially at risk. However, most conservation projects and funds are focused on taxa or

113 habitats generally viewed as charismatic (Ford et al., 2017), but neglect fragile  
114 ecosystems with higher endemism such as cave ecosystems (Mammola et al., 2019a).

115         Because measuring the entire cave biota is challenging and costly, biodiversity  
116 surrogates such as umbrella species can be used to guide targeting of conservation for  
117 large cave communities (Lewandowski et al., 2010; Mammola et al., 2019b; Margules et  
118 al., 2002). Bats make up the second-largest group of mammals with over 1400 described  
119 species distributed across almost all biomes (Burgin et al., 2018; Furey and Racey, 2016).  
120 Within subterranean ecosystems, bats are keystone species by bringing nutrients within  
121 the cave in the form of their guano and excreta after foraging in surrounding areas, thus  
122 providing the basic source of nutrients in these largely lightless zones where the inability  
123 for photosynthesis within the system makes this form of nutrient-transfer critical (Cunha  
124 et al., 2020; Ferreira, 2019). This function makes bats the ideal ecological indicators and  
125 conservation surrogates to inform ecosystem health and priorities to safeguard these  
126 systems (Furey and Racey, 2016; Medellin et al., 2017; Tanalgo et al., 2018). Threats to  
127 cave-bat populations from habitat loss and degradation, hunting and other factors  
128 highlight the need for better methods to develop effective conservation priorities (De  
129 Oliveira et al., 2018; Frick et al., 2019; Furey and Racey, 2016) but a standardised  
130 approach to identify priorities for action to safeguard these systems does not yet exist,  
131 hindering larger-scale prioritisation to protect cave systems and their dependant diversity.

132         Subterranean ecosystems are pervasively threatened by both immediate threats  
133 to the caves themselves, and modifications of the surrounding environment (Phelps et  
134 al., 2016; Tanalgo et al., 2018). Bats are conspicuous in caves, yet large colony sizes  
135 mean that single site impacts can have major consequences on populations especially

136 for locally endemic species (Frick et al., 2019; Sagot and Chaverri, 2015). In addition to  
137 the relatively low reproductive rate in bats that jeopardizes rapid population recovery from  
138 loss and declines (Barclay et al., 2004), bats receive little public support and funding  
139 compared to other large mammals (Fleming and Bateman, 2016). Many caves and karst  
140 habitats are under-protected, for example, only around 13% of the approximately 800,000  
141 km<sup>2</sup> of tropical Southeast Asian karsts are within protected areas (Day and Urich, 2000).  
142 Unprotected karst is especially susceptible to human activities and destruction, for  
143 example in Southeast Asia the average loss is around 5.7% of the area annually due to  
144 mineral mining (Clements et al., 2006; Hughes, 2017; Liew et al., 2016). This loss of bat  
145 cave habitats is coupled with unregulated hunting and tourism, and loss of foraging  
146 habitat; therefore understanding the impacts of these factors on the population status of  
147 bats and biotically important caves are urgently needed (Furey and Racey, 2016; Sedlock  
148 et al., 2014; Tanalgo et al., 2018; Torres-Flores and Santos-Mreno, 2017). It is estimated  
149 that at least 50% of global bat species rely at least partially upon caves (Furey and Racey,  
150 2016) but the degree of threat to bat cave communities and prioritisation has never been  
151 analysed on a global scale (Tanalgo et al., 2018). Understanding only species  
152 distributions, and their extinction risks cannot ensure the protection of all species. The  
153 identification of key spatial areas (e.g., a high number of threatened species and  
154 endemism levels) that require protection will enable conservation biologists to develop  
155 balanced priorities to inform appropriate protection measures. This study is the first  
156 extensive study to explore the global diversity patterns and extinction risk of cave-dwelling  
157 bats and using this information to create an index to guide effective priority making.

158           Here we developed a framework to understand the species-specific and habitat  
159 priorities for cave-dwelling bats to identify hotspots for subterranean conservation by  
160 integrating different facets of biotic importance and risks across different scales. First, we  
161 examined the global (i) patterns of diversity, distribution, and extinction risks of cave-  
162 dwelling bats, and (ii) distribution of species threats. Second, we mapped the broad-scale  
163 and fine-scale priorities of bat caves based on cave biotic potential and vulnerability to  
164 threat. We assess gaps in species information, which is especially important for targeting  
165 appropriate conservation measures in karst environments, which represent one of the  
166 most challenging ecosystems for assessment.

## 167 **Materials and methods**

### 168 **Data source and bat species assessment**

169           We sampled global cave-dwelling bats from two databases, the IUCN Red list (v.  
170 2020.1) and the DarkCideS (v 1.0), a global dataset for cave-dwelling bats, including all  
171 species that occur, use, roost, or hibernate in caves and underground habitats for any  
172 part of their life histories (Data S1-3). All species names were curated and updated using  
173 the Bats of the World: A taxonomic and geographic database (Simmons and Cirranello,  
174 2020). We included species-specific information including species taxonomy, endemism  
175 at geopolitical and biogeographical scales, species range and distributions, conservation  
176 status, population trends, ecological traits, and threatening processes. The habitat  
177 breadth was determined according to the number of habitats a species occurs (Etard et  
178 al., 2020). We used the weighted habitat breadth (%) values in the final analyses. Species  
179 were then classified based on island endemism and country endemism. Whilst country  
180 endemism is not strictly an ecological indicator, it is nonetheless useful as if a species is

181 only present in a single country, then the survival of that species is also subject to  
182 strictures of a single country, which increases vulnerability if protection measures are not  
183 in place. We then compared the species recorded between the IUCN Red list and the  
184 DarkCideS 1.0 to determine gaps in species assessed and cave locality data.

## 185 **Species diversity and distributions**

186 Using IUCN data, we calculated geopolitical (e.g., country or continental) species  
187 richness based on where the species occurs (including native, extant, resident, possibly  
188 extinct or migrant classifications by the IUCN). Second, we analysed and defined  
189 endemism in two ways: geopolitical endemism (i.e., a country endemic) if the species  
190 occurred only in a single sovereign country (Ceballos and Ehrlich, 2002), and island  
191 endemism was classified as island-restricted or predominantly mainland (Faurby et al.,  
192 2018). Species conservation status was assessed according to the IUCN Red list criteria  
193 and were then simplified to obtained binary extinction risk: “Threatened” (e.g., Vulnerable,  
194 Endangered, and Critically Endangered) and “Nonthreatened” (e.g., Least Concern, Near  
195 Threatened). While Data Deficient (DD) species were treated as threatened as they may  
196 face higher or similar threats, hence a lack of data for formal classification (Bland et al.,  
197 2015; Tanalgo et al., 2018; Welch and Beaulieu, 2018). We compared patterns of species  
198 diversity across biogeographical realms (Olson et al., 2001; Olson and Dinerstein, 1998)  
199 (e.g., Indomalayan, Austral-Oceania, Afrotropical, Neotropical, Palearctic, and Nearctic).  
200 Chi-squared test ( $\chi^2$ ) of association was then used to assess the relationship in species  
201 geopolitical endemism, island endemism, conservation status, and population status.

202 We defined country species richness and endemism using simple coherent metrics  
203 which are comprehensible for national or regional level conservation and policymaking

204 (Amori et al., 2011). Using Kendall's  $\tau$  B, we explored the relationships between (1)  
205 country estimated species richness, (ii) % endemic species, and (iii) % threatened to  
206 country land area (km<sup>2</sup>).

### 207 **IUCN-based species extinction risk**

208 The IUCN bat data are imperfect, and many species lack updated assessments,  
209 using the available data we estimated the extinction risks for different groups and species  
210 attributes. Following (Hoffmann et al., 2010; Richman et al., 2015) we estimated the  
211 proportion of threatened species across biogeographical realms, species endemism (e.g.,  
212 geopolitical and island), population trends, trophic levels, and families. We calculated the  
213 proportion of species vulnerable to extinction ( $\hat{p}_{\text{extinction}}$ ) based on the proportion of  
214 threatened species as

$$215 \quad \hat{p}_{\text{extinction}} = (N^{\circ}_{\text{threatened}}) / (N^{\circ}_{\text{species}} - DD_{\text{species}})$$

216 where  $N^{\circ}_{\text{threatened}}$  is the number of threatened species assessed as Vulnerable (VU),  
217 Endangered (EN), and Critically Endangered (CE),  $N^{\circ}_{\text{species}}$  is the total number of species,  
218 and  $DD$  is the number of Data Deficient species, assuming that  $DD$  species will have a  
219 similar extinction risk as of other threatened categories as they may face similar or higher  
220 threats than those that are non-Data Deficient species (Richman et al., 2015; Tanalgo et  
221 al., 2018). We then calculated the lower estimate ( $\hat{p}_{\text{extinction\_lower}} = N^{\circ}_{\text{threatened}} / N^{\circ}_{\text{species}}$ )  
222 with an assumption that  $DD$  species are categorised as non-threatened and upper  
223 estimates ( $\hat{p}_{\text{extinction\_upper}} = (N^{\circ}_{\text{threatened}} + DD) / N^{\circ}_{\text{species}}$ ) with the assumption that  $DD$  is  
224 threatened.

225 Using a Generalized Linear Model (GLM) we determined the predictors of species  
226 extinction risk (threatened vs. nonthreatened) for global species and within each  
227 suborder, Yinpterochiroptera and Yangochiroptera. We used a total of ten explanatory  
228 variables which included geographical variables (geographical range, island endemism,  
229 and geopolitical endemism), biological variables (adult body mass (kg,  $\log_{10}$ ) (Faurby et  
230 al., 2018), generation length (Pacifci et al., 2013), forest dependency (yes or no based  
231 on the IUCN database), weighted habitat breadth (%), and trophic group). We used the  
232 Level-2 habitat data classification from IUCN (2020) that is based on suitable or important  
233 habitats for species (Etard et al., 2020). Weighted habitat breadth was calculated based  
234 on the number of habitat species used divided by all other habitats. For the trophic groups,  
235 we grouped species as frugi-nectarivorous for all species that forage on plant-based  
236 resources (e.g., frugivores and nectarivores). Species that forage on animal resources  
237 (e.g., insectivorous, sanguivorous, species feeding on vertebrates) were grouped as  
238 carnivores, while species that forage on both plant and animal resources were grouped  
239 into omnivores. Our last group of explanatory variables includes the species threat index.

#### 240 **Mapping and quantifying threats**

241 We analysed threats in three different levels: species, cave-site, and country-level.  
242 First, we analysed threats per species based on the simplified expert-based IUCN Red  
243 list. We modified the threat nomenclature by Salafsky et al. (2008) as it does not  
244 completely reflect all threats in bat cave habitats. The Direct threats ( $T_{dir}$ ) refer to the  
245 threats or risks that are directly within or occurring in the cave systems that have  
246 immediate and perceivable impacts on population change or species behaviour. This  
247 threat includes direct human impacts and the use of caves such as harvesting bats,

248 tourism, and mining activities. The Indirect threats ( $T_{ind}$ ) refer to the threats outside or  
249 within cave proximity, which impacts to population is secondary or non-immediate but  
250 detrimental such as deforestation, agriculture, and urbanisation. Lastly, Natural threats  
251 ( $T_{nat}$ ) refers to threats that are natural in origin that may occur and impact directly or  
252 indirectly in the population such as disease and stochastic in nature. We developed the  
253 Species Threat Index ( $STI$ ), which is calculated from the quotient of the sum of species  
254 absolute threat ( $T_{dir, ind, nat}$ ) and the number of threats assessed ( $N^{\circ} T$ ). We compared  $STI$   
255 ( $STI_{dir, ind, nat}$ ) across the biogeographical realm, endemism, conservation status, and  
256 population trend using a non-parametric Kruskal-Wallis test.

$$257 \quad (STI_{species} = \sum T / N^{\circ} T)$$

258 We constructed a separate generalized linear model (GLMs) with a binomial  
259 distribution (logit link function) in JAMOVI version 2, using GAMLj module (Gallucci, 2019)  
260 to understand the link between species traits to a certain threatening process. We used  
261 adult body mass (g,  $\log_{10}$ ), geographical range ( $\log_{10}$ ) and trophic levels as the predictor  
262 variable and threat status (threatened or not threatened) as response variable (Atwood  
263 et al., 2020). We only modelled threatening processes that impacted at least 10% of the  
264 cave-dwelling bats.

265 Second, we used remotely sensed data to measure cave vulnerability to threats.  
266 For landscape features, we included canopy cover height (Simard et al., 2011), tree  
267 density (Crowther et al., 2015), distance to bodies of water (Yamazaki et al., 2015), bare  
268 ground cover change (Song et al., 2018), short vegetation cover change (Song et al.,  
269 2018), tall tree cover change (Song et al., 2018) and for vulnerabilities we include distance  
270 to urban areas (Song et al., 2018), distance to roads (Meijer et al., 2018), mine density

271 (Labay et al., 2017), night light (Earth at Night, 2019), relative pesticide exposure (Maggi  
272 et al., 2019), and population density (Hughes, 2019; SEDAC, 2020). We then assessed  
273 the correlation between landscape variables to cave species biotic scores and the extent  
274 was compared across biogeographical realms.

275 Third, we ranked each country based on total species richness, the proportion of  
276 both threatened and endemic species and its concordance to sociodemographic and  
277 environmental indicators as rudimentary indicators of a country's resources to  
278 representing its capacity to monitor and protect its species and environment (e.g.,  
279 Convention of Biological Diversity) (Amori et al., 2011; McGeoch et al., 2010). In our  
280 analyses, we used percent forest cover (The World Bank, 2020) as a proxy for habitat  
281 intactness and quality per country, with the assumption that caves surrounded by intact  
282 forest are more protected and least vulnerable to intrusions (Cajaiba et al., 2021; Phelps  
283 et al., 2018; Tanalgo et al., 2018). While we used country cement production (TTM\*1000)  
284 (Index Mundi, 2020) as a proxy to species susceptibility to direct threats assuming that in  
285 areas with higher cement production species are more prone to habitat loss and direct  
286 disturbances (Liew et al., 2016; Tanalgo et al., 2018). We applied Kendall's  $\tau B$  to test for  
287 the ranked-concordance in (i) country-level species richness (log 10), (ii) % endemism,  
288 (iii) % threatened species, (iv) percent forest cover (%), log 10 (iii) socioeconomic  
289 indicators: GDP per capita (log 10), country cement production (TTM). We mapped and  
290 visualised estimated species richness, endemism, and threatened species using QGIS  
291 version 3.14.

## 292 **Global Bat Cave Vulnerability Index (BCVI-G)**

293 Using the global ‘Bat Cave Vulnerability Index’(Tanalgo et al., 2018) (hereafter  
294 ‘BCVI-G’), we assessed habitat-level priorities of pooled bat caves (Data S3). The  
295 prioritisation index is based on two integrated sub-indices representing cave biotic  
296 potential and vulnerability. The first component of the index is the cave Biotic Potential  
297 (BP), which analyse, and measure bat cave biological values based on bat species  
298 richness and attribute status. The second component is the cave Biotic Vulnerability (BV)  
299 measures the cave landscape feature and vulnerability to threats. Using the index, we  
300 mapped and constructed broad-scale and fine-scale cave priorities representing biome  
301 and site-level analyses respectively. We modified the first version of BCVI to address  
302 some of its limitations for incomplete or unstandardised data and make it relevant for a  
303 comparative global-scale prioritisation scheme for the new BCVI-G.

304 
$$BCVI-G = BP \times BV$$

305 **Cave Biotic Potential (BP)**

306 We excluded data from population estimates to minimise bias from the missing  
307 and unstandardised assessments in caves datasets, and which is a component of the  
308 first BCVI and can be used at more local scales. Second, in addition to species attribute  
309 status scoring (conservation status, population trends, country endemism, and island  
310 endemism (Table 1), we incorporated evolutionary units including evolutionary  
311 distinctiveness (ED, based on EDGE score (Isaac et al., 2007)) and corrected weighted  
312 endemism (CWE) to factor the cave biotic potential (BP). We calculated corrected  
313 weighted endemism (CWE) (Crisp et al., 2001; Laffan and Crisp, 2003) by dividing  
314 weighted endemism (WE) by the total cave species richness (e.g., absolute counts of

315 species). This measures the proportion of bat species in a cave site. We then calculated  
 316 BP using the equation

317 
$$BP_{cave\ x} = \sum S_{cave\ x} / \text{Max.} \sum S_{cave\ y}$$

318 where cave BP is the calculated quotient of the sum of cave species attributes scores ( $\sum$   
 319  $S_{cave\ x}$ ) and the highest maximum ( $\sum S_{cave\ y}$ ) sum of species attribute scores from all  
 320 sampled caves ( $cave\ y$ ) within the single site or the entire biome. Where  $S_{cave\ x}$  is the sum  
 321 score of  $n^{th}$  bat cave species evolutionary distinctiveness (ED), corrected weighted  
 322 endemism ( $\times 100$ ) (CWE), conservation status (Cons), population trends (Pops), country  
 323 endemism (E), and island endemism (Isl).

324 
$$S_{cave\ x} = \sum \text{Species 1 (ED + CWE (\times 100) + Cons + Pops + E) + Species 2 (ED + CWE}$$
  
 325 
$$(\times 100) + Cons + Pops + E) + \text{Species 3 (ED + CWE (\times 100) + Cons + Pops + E)}$$

326 The  $BP_{cave}$  index score ranges from 0.00 to 1.00, with values near 1.00 indicates  
 327 higher cave biotic potential and scaled to four levels of priority (Table 2).

328 **Table 1.** Scoring of species attributes for calculation of cave Biotic Potential (Bp) based  
 329 on species Population trends (pops), Conservation status (cons), Country endemism  
 330 (E), and Island endemism (Isl).

pops	score	cons	score	E	score	Isl	score
Increasing	2.5	Least concern	2	Non-endemic	5	Non-islandic	5
Stable	5	Near Threatened	4	Endemic	10	Islandic	10
Unknown	7.5	Vulnerable	6				
Decreasing	10	Endangered	8				
		Critically endangered/Data deficient	10				

331  
 332

333 **Table 2.** Cave Biotic Potential (Bp) priority status and description of the probable  
 334 scenario. The interpretation of the probable scenario was based on Tanalgo et al. 2018.

<b>Bp cave<sub>x</sub>/BP cave<sub>max</sub></b>	<b>BP status</b>	<b>Probable relative scenario</b>
0.00 – 0.24	Level 4	Contain the <i>lowest</i> species richness, <i>lowest</i> sum of species attributes, with the absence of threatened and endemic species. Probably no rare species in the cave.
0.25 – 0.49	Level 3	Species richness and the sum of species attributes may be lower. Threatened and endemic species may be present. Probably no rare species in the cave.
0.50 – 0.74	Level 2	Cave may be high in species richness and sum species attributes, with the presence of few endemic and threatened species. Rare species may potentially present.
0.75 – 1.00	Level 1	Highly species richness, with <i>highest</i> sum of species attributes. Several threatened and endemic species is present. Cave contains the rarest species.

335

336 **Mapping landscape features and Cave Vulnerability (BV)**

337       The computed cave BP is then complimented with the cave Biotic Vulnerability  
 338 (BV) to derive the final cave alphanumeric priority based on vulnerability. First, we  
 339 mapped and measured the extent of geophysical features represented landscape  
 340 features and threats in a single cave following (Hughes, 2019) using ArcGIS version 10.3  
 341 (see reference for detailed methods for every landscape feature). Because of the  
 342 relatedness of landscape features to represent cave vulnerability to threats, we selected  
 343 representative features to enter in BV calculations: (i) distance to urban area, (ii) distance  
 344 to roads, (iii) tree density, (iv) canopy cover, (v) mining density, and (vi) distance to bodies  
 345 of water (for arid biomes). We calculated cave Biotic Vulnerability (BV<sub>cave</sub>) as the quotient  
 346 of summed scores of *NT* and the total number of geophysical features assessed (*N*<sup>o</sup>, *N*<sup>o</sup>  
 347 = 5).

348

$$BV = \sum NT/N^o$$

349 where  $NT$  ( $NT = T_x / T_{max} y$ ) is the score of geophysical or landscape features ( $T$ ) based  
 350 on the quotient between the measured extent of geophysical features in a specific cave  
 351 ( $T_x$ ) divided to the maximum value of measured extent of threats and landscape features  
 352 in all sampled caves ( $T_{max} y$ ). The  $N$  score ranges from 0.00 to 1.00 and is scaled in four-  
 353 level range scoring representing cave relative probable scenario (Table 3).

354 **Table 3.** Scoring system for Biotic Vulnerability (BV) index based on the extend of the  
 355 landscape features and cave probable relative scenario.

Threat or landscape features	$T_x / T_{max} y$	BV score	Probable relative scenario (worst to best)
<b>Distance to city/urbanisation</b>	0.00 - 0.24	1	Cave site is within or surrounded by heavy urbanisation; highly vulnerable to visitation and unregulated tourism.
	0.25 - 0.49	2	
	0.50 - 0.74	3	
	0.75 - 1.00	4	
<b>Distance to roads</b>	0.00 - 0.24	1	Cave is not surrounded by urbanisation, or very far any urbanisation. Cave is probably located in a pristine area. Cave site is accessible by roads or human trails. Highly accessible and vulnerable to visitation and intrusion. Cave sites with hyper-population may be vulnerable to hunting.
	0.25 - 0.49	2	
	0.50 - 0.74	3	
	0.75 - 1.00	4	
<b>Mine density</b>	0.75 - 1.00	1	Cave site is inaccessible by roads or human trails. Roads may be absent. Difficult to access by human visitation. Cave site is in a high mining density area.

			Extensive mineral extraction occurs within the cave site.
	0.50 - 0.74	2	
	0.25 - 0.49	3	
	0.00 - 0.24	4	Mining is low or probably absent.
<b>Canopy cover</b>	0.00 - 0.24	1	Cave site is in a more open area. The occurrence of deforestation is probably high. The cave is potentially in an arid environment with low tall tree vegetation. The cave is very exposed and highly vulnerable to human intrusion and other disturbances.
	0.25 - 0.49	2	
	0.50 - 0.74	3	
	0.75 - 1.00	4	Cave is probably in a pristine or robust forested area. Tall tree vegetation is probably high. Foraging grounds for bats are highly available. The cave is probably located in a protected area.
<b>Tree density</b>	0.00 - 0.24	1	Cave site is in a heavily disturbed area. Deforestation and other surface threats (e.g., mining and urbanisation) may be highly present. The cave site is very exposed and highly vulnerable to population loss.
	0.25 - 0.49	2	
	0.50 - 0.74	3	
	0.75 - 1.00	4	Cave site occurs in a very pristine and intact forested area. Surface threats may be low or absent. Foraging grounds for bats are highly available. The cave is

probably covered by  
protected areas.

356

357 **Mapping conservation priorities**

358 We aimed to assess the vulnerability of global bat caves based on community level  
359 bat data and mappable landscape features and threats (Figure 7). We performed BCVI-  
360 G prioritisation at a number of spatial scales. First, the broad-scale priorities represent  
361 biome-dependent analyses. Secondly, we measured at the site-level diversity and  
362 vulnerability to encompass the fine-scale prioritisation. Here we set cave priorities at a  
363 national scale, with the assumption that priorities should be comparable to guide decision  
364 making at any scale. The alphanumeric index derived from BCVI-G is divided into four  
365 priority categories (Table 4). We then compared the mean biotic vulnerability (i.e., values  
366 range from 1 to 4, the lower the values the higher the vulnerabilities) versus cave biotic  
367 potential status, and then the priorities from both scales across biogeographical realms  
368 and biomes using a chi-squared test ( $\chi^2$ ). We used Pielou's index in PAST - to assess  
369 evenness in BCVI-G and priorities between scales.

370 **Table 4.** Priority scales of caves based on Bat Cave Vulnerability Index-Global (BCVI-  
371 G).

BCVI-G	Priority-level	Condition	Potential action
1A, 1B, 2A	Red Priority caves	High diversity and high threat exposure	Cave needs immediate action
1D, 1C	Green Priority caves	High diversity, but threat exposure is absent	Cave needs monitoring
2B, 2C, 2D, 3A, 3B, 3C, 3D	Yellow Priority caves	Under high threat, and moderate diversity. May have already lost species.	Cave needs intervention or may no longer be worthwhile

4A, 4B, 4C, 4D	Blue Priority caves	High to low threat, with very low diversity. May have already lost species.	Cave not worth attention
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372

373 **Results**

374 **Global diversity and distribution of cave-dwelling bats**

375 The IUCN lists a total of 679 ( $N_{cave} = 679$  spp./1400) cave-dwelling species  
376 constituting the 48.5 % of described global bat species belonging to all bat families.  
377 Vespertilionidae comprised the largest proportion of all cave-dwelling species with 215  
378 species ( $\%_{cave} = 32$ ,  $\%_{global} = 43$ ), and followed by Phyllostomatidae ( $N = 96$ ,  $\%_{cave} = 14$ ,  
379  $\%_{global} = 45$ ) (Table S1). Most species are concentrated in the tropical regions ( $\chi^2 = 205.83$ ,  
380  $df = 5$ ,  $P < 0.0001$ ) with 30% ( $N = 227$  spp.) of global cave-dwelling bats in the  
381 Indomalayan region (Figure 1). At a country level, Indonesia has the highest number of  
382 species ( $N = 104$  spp.), followed by China ( $N = 98$  spp.), and India ( $N = 82$  spp.) (Figure  
383 S1; Data S2). In addition, Indonesia ( $N = 18$  spp.,  $\%_{threatened} = 17\%$ ), and India ( $N = 10$   
384 spp.,  $\%_{threatened} = 12\%$ ), were the countries with highest number of threatened species.  
385 Unsurprisingly, we found congruence between country land area ( $km^2$ ) and estimated  
386 species richness ( $\tau = 0.40$ ,  $P < 0.001$ ),  $\%_{threatened}$  ( $\tau = 0.41$ ,  $P < 0.001$ ) and  $\%$   
387 endemism ( $\tau = 0.19$ ,  $P < 0.001$ ).

388 Globally, 32% of cave-dwelling bat species were endemic to a single country and  
389 23% to small islands. Most of the geopolitically endemic species are classified as  
390 threatened (63.58%,  $N = 110$  spp.) compared to less threatened (21.74%,  $N = 63$  spp.)  
391 ( $\chi^2 = 120.50$ ,  $df = 5$ ,  $P < 0.0001$ ). The majority of data deficient species (57%,  $N = 47$  spp.)

392 are country endemic. Moreover, the distribution of island endemic species significantly  
393 differed across conservation statuses ( $\chi^2 = 192$ ,  $df = 2$ ,  $P < 0.001$ ). The 77% ( $N = 520$   
394 spp.) of the species are found on mainland and near-shore islands and 23% ( $N = 159$   
395 spp.) are restricted to islands (Figure 1). Unsurprisingly, 75% of island species are country  
396 endemic versus 81% of the non-endemic occur in mainland areas ( $\chi^2 = 171$ ,  $df = 1$ ,  $P <$   
397  $0.001$ ). Moreover, the proportion of species in threatened categories within island  
398 endemism is higher for island restricted species (40%;  $N = 43$  spp.) compared to mainland  
399 species (21%;  $N = 173$  spp.) ( $\chi^2 = 35.4$ ,  $df = 2$ ,  $P < 0.001$ ).

#### 400 **Patterns of threat and extinction risks**

401 Using the IUCN data, the estimated extinction risks of cave-dwelling bat species  
402 ( $\hat{\rho}_{\text{extinction}} = 15\%$ , 13-25%) is lower than that of all bat species ( $\hat{\rho}_{\text{extinction}} = 20\%$ , 16-35%),  
403 of which 37% are in a threatened category and 12% are data deficient (Figure 2A, Table  
404 S2). Extinction risk is relatively higher when compared across sub-orders, families, and  
405 trophic groups compared to all species. Moreover, country endemic species ( $\hat{\rho}_{\text{extinction}} =$   
406  $36\%$ , 29-50%) and species occurring on islands (islandic:  $\hat{\rho}_{\text{extinction}} = 40\%$ , 35-48%) are  
407 facing exceedingly high extinction risk (Figure 2A, Table S2). Additionally, a strong link  
408 between narrow geographic range and species extinction risk can be attributed in all  
409 cave-dwelling species ( $\beta = -1.94$ ,  $P < 0.001$ ) and for both suborder models (Yin:  $\beta = -$   
410  $2.34$ ,  $P < 0.001$ ; Yang:  $\beta = -2.02$ ,  $P < 0.001$ ). Country endemism ( $\beta = -0.67$ ,  $P = 0.013$ )  
411 and island endemism ( $\beta = 1.00$ ,  $P < 0.001$ ) are also linked to extinction risk globally. But  
412 varies within suborders, island endemism could only predict species with a high extinction  
413 risk from Yangochiroptera ( $\beta = 1.17$ ,  $P < 0.001$ ) and geopolitical only for  
414 Yinpterochiroptera ( $\beta = -1.19$ ,  $P = 0.028$ ) (Figure 2B, Table S3). Moreover, none of the

415 biotic variables (trophic level, generation length, and adult body mass) included in the  
416 model could predict extinction risk globally and within suborders. Among threat variables,  
417 only direct threats showed significant association globally ( $\beta = 2.221$ ,  $P = 0.008$ ) and  
418 between suborders (Yin:  $\beta = 2.84$ ,  $P = 0.043$ ; Yang:  $\beta = 2.50$ ,  $P = 0.031$ ) (Figure 2B).

419         Nearly three-quarters (69%,  $N = 466$  spp.) of the cave-dwelling bat species are  
420 exposed to various threats according to the IUCN. The proportion of direct (Kruskal-Wallis  
421 test:  $\chi^2 = 13.02$   $df = 5$ ,  $P = 0.02$ ) and indirect (Kruskal-Wallis test:  $\chi^2 = 30.10$ ,  $df = 5$ ,  $P <$   
422  $0.01$ ) threats differed regionally. However, there is a large disparity in the species  
423 threatened by natural threats with only 10% ( $N = 70$  spp.) ( $STI_{nat} = 1.40$ ) of the species  
424 assessed facing geo-climatic induced threats. (Figure S1, Table S4). Species risks to  
425 dominant threats such as agricultural conversion ( $\beta = -0.214$ ,  $P = 0.017$ ) and deforestation  
426 ( $\beta = -0.154$ ,  $P = 0.05$ ) were significantly linked to small geographic range. Conversely,  
427 species with large geographic ranges were more at risk from pollution ( $\beta = 0.498$ ,  $P =$   
428  $0.003$ ). Species adult body mass is a strong predictor of species vulnerability for hunting  
429 and bushmeat for large species ( $\beta = 0.845$ ,  $P < 0.001$ ). Within trophic levels, omnivores  
430 have a significantly higher risk from mining and quarrying compared to frugi-nectarivores  
431 ( $\beta = -1.029$ ,  $P = 0.023$ ) and carnivores ( $\beta = -0.636$ ,  $P = 0.026$ ). Frugi-nectarivores are  
432 more vulnerable to hunting and harvesting ( $\beta = 2.256$ ,  $P = 0.001$ ), and insectivores to  
433 pollution ( $\beta = 1.483$ ,  $P = 0.044$ ) compared to other trophic levels (Figure S2 and Table  
434 S5).

435         On a global scale, seven out of twelve variables showed a significant relationship  
436 with summed species biotic scores ( $\Sigma S_{cave\ x}$ ) but none of the landscape features and  
437 threat variables showed a strong correlation to diversity (Figure 4A). Tree density, bare

438 ground cover change, and short vegetation cover change showed a positive correlation  
439 with bat cave diversity. Conversely, distance to bodies of water, tall tree loss, nightlight,  
440 and population density showed a significant negative correlation though this may in part  
441 reflect the challenges of sampling caves in tall, forested areas, but also highlights that  
442 recently disturbed areas may be threatened. Furthermore, socioecological variables,  
443 GDP per capita, % forest cover, and cement production as a proxy to assess the  
444 vulnerability of geopolitically endemic species showed consistent correlations amongst  
445 species diversity attributes (Figure 3C, Data S2).

#### 446 **Habitat priorities for cave-dwelling bats**

447 The degree of evolutionary distinctiveness (ED) constructed on bat cave data  
448 significantly differed regionally (Kruskal-Wallis,  $\chi^2 = 1615.65$ ,  $df = 5$ ,  $P < 0.001$ ). The  
449 highest ED was in the Neotropical region ( $ED_{\text{mean}} = 11.92$ ) with the lowest in the Palearctic  
450 ( $ED_{\text{mean}} = 5.54$ ). Cave weighted endemism is highest in the Austral-Oceania ( $CWE_{\text{mean}} =$   
451  $6.32$ ) consistent with the proportion of species endemism observed in the region (Kruskal-  
452 Wallis,  $\chi^2 = 1584.84$ ,  $df = 5$ ,  $P < 0.01$ ) (Figure 4). Prioritisation at broad and fine scales  
453 showed that the majority of cave sites are highly vulnerable to threats (Figure 5A-D). In  
454 fine-scale analysis, we found a significant relationship in biotic potential and biotic  
455 vulnerability score (Kruskal-Wallis test:  $\chi^2 = 14$ ,  $df = 3$ ,  $P = 0.003$ ). Caves with higher  
456 biotic potential are linked to lower vulnerability caves (i.e., cave systems located in  
457 relatively pristine ecosystems), while there is a lower biotic potential in caves with higher  
458 threats and vulnerability (Kruskal-Wallis test:  $\chi^2 = 6.45$ ,  $df = 3$ ,  $P = 0.092$ ). This pattern is  
459 consistent at broad scales, but the relationship was not significant (Kruskal-Wallis test:  $\chi^2$

460 = 6.45,  $df = 3$ ,  $P = 0.092$ ) (Figure 5E-F), likely relating to the lack of site-specific data  
461 needed.

462 Indices derived from the site-level and biome-level BCVI-G were used to construct  
463 a comparative prioritisation across the biogeographical realm and climate regions (Data  
464 S3, Table S6). On a broad scale, 95% of the caves show high biotic vulnerability (Status  
465 A) but 88% of the caves have a lower biotic potential level (Level 4) in contrast to only 1%  
466 with high biotic values (Level 1) (Figure 6A). The integration of two sub-indices within the  
467 broad-scale equates to the uneven and high proportion (83%) of “4A” lower vulnerability  
468 index values for overall sampled caves ( $J' = 0.378$ ) (Figure 6A). Conversely, fine-scaled  
469 BCVI-G analyses showed a more even distribution of indices values ( $J' = 0.917$ )  
470 compared to the broad-scale analyses. At a fine scale, there is an increase at 45% in the  
471 proportion of caves with high biotic potential (Level 1), 14% at a mid-high level (Level 2),  
472 and 41% of caves had a lower biotic potential (Level 4). Subsequently, an even  
473 distribution of biotic vulnerability was also observed with fine-scale BCVI-G, 45% and 41%  
474 of the caves are in high (Status A) to mid-high (Status B) vulnerability, respectively.  
475 Moreover, 10% of the caves are in the “1A” high vulnerability index. A high proportion of  
476 high vulnerability caves occur in tropical realms (Figure 6A). We found a significant  
477 difference in the cave vulnerability index in both scales (Broad:  $\chi^2 = 8977.87$ ,  $df = 7$ ,  $P <$   
478  $0.001$ ; Fine:  $\chi^2 = 1303.09$ ,  $df = 15$ ,  $P < 0.001$ ).

479

480 **Table 5.** Comparison of achieved per cent priorities between broad-scale and fine-scale  
481 analyses using the Bat Cave Vulnerability Index- Global (BCVI-G). Full data of BCVI-G  
482 results and priorities in two scales are in Supplementary Data 3 and Supplementary  
483 Table 6.

Priority-level	Broad-scale	Fine-scale	Global average	Priority difference
Red	3%	28%	15%	+25%
Green	0%	4%	2%	+4%
Yellow	9%	36%	23%	+27%
Blue	88%	32%	60%	-56%

484  
485 Our analysis of the sampled caves showed that the priority levels of caves are  
486 scale-dependent and varied significantly across spatial scales ( $\chi^2 = 1281.43$ ,  $df = 3$ ,  $P <$   
487  $0.001$ ) (Figure 6B; Table S6). Caves were classified on the basis of the need for different  
488 types of intervention or management based on threat and biotic characteristics, most  
489 importantly these include “red caves”, which whilst diverse are threatened and in need of  
490 intervention, and “green caves” which are also diverse but not currently at risk, whilst  
491 “yellow caves” are at an intermediate level in terms of diversity and may need intervention  
492 to prevent species loss and allow recovery. On a broad scale, only 3% of the sampled  
493 caves are at “Red Priority”, 9% at “Yellow Priority”, and 88% at “Blue Priority” levels. The  
494 low proportion of high priority caves are significantly concentrated in the Neotropical  
495 (45%), Afrotropical (18%) and Indomalayan (16%) regions ( $\chi^2 = 204.20$ ,  $df = 10$ ,  $P <$   
496  $0.001$ ) (Figure 6B, Table S6). While on a fine scale, there is a significant increase in the  
497 proportion and evenness ( $J' = 0.395$ ) of high priority caves compared to broad-scale. Of  
498 the sampled caves, 28% are “Red Priority”, 36% are “Yellow Priority”, and 4% are “Green  
499 Priority” caves that host high biotic potential but low vulnerability (Figure 6B, Table 5).  
500 The concentration of high priority caves is highest in the Palearctic (30%) closely followed  
501 by Neotropical (29%) and Indomalayan (28%) ( $\chi^2 = 73.93$ ,  $df = 15$ ,  $P < 0.001$ ). When  
502 compared by biomes and climatic regions, the 58% and 25% of high priority caves are

503 concentrated in the tropics and temperate regions respectively ( $\chi^2 = 56.76$ ,  $df = 12$ ,  $P <$   
504  $0.001$ ) (Figure 6B).

## 505 **Discussion**

### 506 **Bats and caves in the changing world**

507 This is the first study to present the integrative taxonomic and habitat-level  
508 conservation priorities for cave-dwelling bats on a global scale. In our previous work  
509 (Tanalgo et al., 2018), we developed BVCI for local and community-level applications,  
510 including local factors such as population and other site-specific factors. However, whilst  
511 such an approach is ideal for a region or country, especially where a single team can  
512 inventory all sites, priorities across a broader region require more data that can be collated  
513 remotely and can be sourced from multiple teams. Thus the new approach provides an  
514 index that can be applied across wide regions, or as here globally, and like the Essential  
515 Biodiversity Variables (EBV's) (Jetz et al., 2019) allows remotely sensed data to be  
516 integrated with on the ground data to provide a more robust index for prioritisation. Caves  
517 and underground habitats are used by almost half (48.5%,  $N = 679$  spp.) of all bat species,  
518 with a large fraction restricted to small ranging islands (23%) and considered endemic  
519 (32%) or threatened (25%). We observed higher extinction risk in species with a narrow  
520 geographical range distribution (e.g. island and nationally endemic species), consistent  
521 with other studies (Jones et al., 2003; Welch and Beaulieu, 2018). The association of  
522 endemism level to extinction risk varies phylogenetically showing that closely related  
523 species have a similar association (Jones et al., 2003), island and geopolitical endemism  
524 is respectively correlated to suborders Yangochiroptera and Yinpterochiroptera. Whilst  
525 cave-dwelling bats are not the sole biological indicators in subterranean ecosystems, their

526 diversity may offer a relatively cost-effective conservation surrogate for systematic  
527 monitoring to protect the vast diversity of cave-dependent species. The patterns of cave  
528 bat diversity and distribution are consistent with the patterns observed for global bats,  
529 peaking in the tropics and particularly in the Indomalayan, Afrotropical, southern  
530 Palearctic, and Neotropical regions (Burgin et al., 2018; Frick et al., 2019). However  
531 estimates of diversity and proportion of threatened species are likely to be underestimated  
532 due to current taxonomic gaps, large numbers of undescribed cryptic species, and lack  
533 of accurate species distributions assessments for global bats (Francis et al., 2010; Murray  
534 et al., 2012; Welch and Beaulieu, 2018).

### 535 **Understanding threats to species and habitats**

536 Direct threats showed a strong link to species extinction risks in all our models.  
537 Large colonies in many cave-dwelling bat species make them more vulnerable to direct  
538 anthropogenic disturbances such as hunting, harvesting, and unregulated tourism (Sagot  
539 and Chaverri, 2015). Large-bodied species are likely vulnerable to hunting and  
540 harvesting. The 18% of cave-dwelling bats are threatened by hunting, largely in parts of  
541 the Afrotropical and Indomalayan regions. This represents a large proportion (62%) of bat  
542 species (167 spp.) hunted globally (Mildenstein et al., 2016), and is likely to be an  
543 underestimate as hunting in many regions is poorly documented. The high level of hunting  
544 and harvesting in the Old-World tropics is primarily driven by subsistence and primarily  
545 localised particularly in areas with high levels of poverty and driven by the demand for  
546 protein sources, food, and traditional medicine (Cardiff et al., 2009; Goodman, 2006;  
547 Mickleburgh et al., 2009). The most frequently hunted species are common and hyper-  
548 abundant cave species (e.g., *Rousettus amplexicaudatus*, *Eonycteris spelaea*, *Eidolon*

549 *helvum*), which have a smaller portion of their range protected because of their  
550 widespread, non-threatened status (Aziz et al., 2021). However, as a consequence of the  
551 absence of statutory protection for many common species intensifies hunting and can  
552 cause even common and abundant species to become locally or regionally extinct  
553 (Tanalgo and Hughes, 2019). Separately, unregulated tourism is a direct threat to 38% of  
554 the cave-dwelling species and unregulated activities may alter cave microclimate and  
555 affect sensitive species.

556 Cave-dwelling bats are also at risk from land degradation as a result of  
557 deforestation and agricultural conversions (Cajaiba et al., 2021; Jones et al., 2009). Over  
558 50% of the species are already losing habitat in close proximity to their roost sites but this  
559 may be underestimated as at least 12% of the species are data deficient and even more  
560 species lack updated red list assessments. Disturbed caves in deforested and agricultural  
561 lands drive the loss of specialist bats, whereas fewer disturbed caves support high  
562 species richness and abundance (Cajaiba et al., 2021). Furthermore, increased  
563 deforestation and vegetation removal around cave sites increases the exposure of caves  
564 to human intrusion and potentially increases vulnerability to direct threats e.g., hunting  
565 and tourism. Additionally, extractive industries of mining and quarrying threaten more than  
566 a quarter of cave-dwelling bats, through degradation and destruction of caves and  
567 alterations of surface vegetation, which creates barriers that limit species movements and  
568 foraging (Theobald et al., 2020). Although our analyses showed that average mining  
569 density is higher in the Nearctic region, this largely omits quarrying for limestone (as such  
570 maps are rarely available)(Sonter et al., 2018). However, cement export is significantly  
571 higher in countries with high cave-bat species richness and high numbers of threatened

572 species (e.g., in Southern China and throughout mainland Southeast Asia), in part  
573 because of the extensive limestone karsts in these regions.

#### 574 **Habitat conservation and priorities**

575 Globally, 3% to 28% of bat caves need immediate conservation interventions,  
576 while 9% to 35% require monitoring due to high diversity, but low risk at present. Overall,  
577 the patterns of habitat level priorities are consistent with previous global studies  
578 comparing the value of broad- and fine-scale analyses in identifying priorities. The broad-  
579 scale analysis provides an overview of patterns to guide where further focus and fine-  
580 scale prioritisation is needed. There is a slight overlap in priorities between scales such  
581 as some high priority caves from the broad-scale analysis are the same in the fine-scale  
582 prioritisation, but not all. Broad-scale prioritisation if applied at regional or biome levels  
583 may facilitate the identification of priority caves. However, we found broad-scale  
584 measures underestimated the priorities of highly vulnerable bat caves (i.e., less even  
585 distribution of priorities) with a 25% difference with fine-scale prioritisation (Figure 6, Table  
586 5). In addition, fine-scale prioritisation (e.g., national-level priorities) enables community-  
587 level interactions and responses to be encompassed and also accounts for rare species  
588 and the impacts of threats on local populations (Cajaiba et al., 2021; Tanalgo et al., 2018),  
589 highlighting the need for good monitoring and assessment data as a basis for priority  
590 setting. Furthermore, context-specific threats (e.g., vulnerability to religious activities in  
591 Buddhist regions, where caves often become temples or religious sites) need to be  
592 accounted for explicitly for indices to be effective.

593 The integration of vulnerability to effective conservation planning is often  
594 challenging due to vague or inconsistent approaches and definition (Sarkar et al., 2006).

595 While most habitat protection centres on taxonomic diversity (e.g., counts, abundance,  
596 and rarity) (Brum et al., 2017; Hartley and Kunin, 2003). Common species can provide  
597 ecological services (e.g., nectarivorous cave bats as pollinators, insectivores for pest  
598 control) and the reduction of population counts may indicate habitat health due to changes  
599 in service provision. Whereas, species are often prioritised based on their combined  
600 distinct ecological function, high sensitivity and vulnerability to declines (Mouillot et al.,  
601 2013). Within cave ecosystems maintaining the functioning of the subterranean  
602 ecosystems requires maintaining both diversity and abundance (Bregović et al., 2019;  
603 Phelps et al., 2016; Tanalgo et al., 2018), thus a holistic tool that incorporates diversity,  
604 rarity and function is needed as a basis for conservation-decision making. Numerous  
605 indices based on different dimensions and attributes of cave biodiversity have been  
606 developed but few integrate landscape features, and greater standardisation is needed.  
607 Whilst various taxa have been used as indicators for diversity in subterranean habitats  
608 (Souza Silva et al., 2015), bats represent holistic surrogates for cave conservation  
609 because they not only provide the main source of energy for cave ecosystems but are  
610 also easier to assess and reflect changes from both internal and surfaces (Cajaiba et al.,  
611 2021; Jones et al., 2009). Our vulnerability index includes not only species diversity  
612 measures (Furman and Özgül, 2004; Niu et al., 2007) but also encompasses evolutionary  
613 distinctiveness and threat exposure of both species and its habitat (Cajaiba et al., 2021;  
614 Phelps et al., 2016; Tanalgo et al., 2018). The inclusion of these measures enables  
615 prioritisation of cave habitats with rare and higher functional diversity attributes (Jetz et  
616 al., 2014; Srivastava et al., 2012) complimenting the metrics based on geopolitical  
617 endemism and conservation status from IUCN (Isaac et al., 2007; Jetz et al., 2014;

618 Martín-López et al., 2009) which are commonly used within prioritisation schemes. While  
619 the expert-based Red list developed by the IUCN is the most comprehensive basis for  
620 conservation and species protection, it is not free from biases (Hughes et al., 2021;  
621 Martín-López et al., 2011; Trimble and Aarde, 2010), especially for bats, in which a large  
622 proportion of species are either taxonomically and spatially under-sampled particularly in  
623 most megadiverse and developing countries (Zamora-Gutierrez et al., 2019).

624 For an index to be effective, a clear understanding of diversity patterns and  
625 priorities at national levels is an essential first step to implementable policy targets (Doi  
626 and Takahara, 2016; Rudd et al., 2011) (Figure 7). Additionally, few countries have any  
627 policy related to the protection of caves and their biota (Whitten, 2009). For example, in  
628 the Philippines, the National Cave Conservation Committee aims to identify caves for  
629 protection has very broad criteria and focused on archaeological and touristic values  
630 rather than ecological components which hamper effective protection or priority setting.  
631 National Biodiversity Action Plans (NBSAPS) should include standard provisions for  
632 priority identification and monitoring which include all habitats (Martín-López et al., 2009),  
633 thus frameworks such as this can provide information that is both consistent between  
634 countries and can be usefully applied at national levels.

### 635 **Caveats and opportunities for bat cave conservation in the Anthropocene**

636 The uneven distribution of threats and a lack of understanding of their impacts  
637 remains a challenge for global bat conservation. Developing any index requires  
638 pragmatism in finding indicators that are reliable enough, but for which there is sufficient  
639 data available. For example, the interpretation of extinction risk requires caution as factors  
640 may act in synergy and we could not fully account for the intensity of human-induced

641 threats (Howard et al., 2019). This assessing vulnerability is particularly challenging to  
642 identify in the context of cave biota, in which even a single disturbance may alter the  
643 entire sensitive biota and ecosystem, yet causal drivers are challenging, and other  
644 proximal indicators (e.g. accessibility) must be used as indicators (Cajaiba et al., 2021;  
645 De Oliveira et al., 2018; Phelps et al., 2018). Furthermore, the degree of expertise  
646 required for bat and cave studies means less data is available compared to other  
647 taxonomic groups (Herkt et al., 2017; Zamora-Gutierrez et al., 2019). For instance, in our  
648 cave prioritisation, we only accounted the 59% of the global cave-dwelling species, and  
649 species coverage varied by region, for example, Indonesia has some of the highest  
650 estimated bat cave species richness yet its contribution to the dataset based on surveys  
651 and assessments is among the lowest. In addition, cave community data is lacking for  
652 biodiverse regions such as from the Afrotropical and Indomalayan regions. Furthermore,  
653 accurate systematic and taxonomic studies for bats are vital to appropriate conservation  
654 as caves host high endemism and many cave bats (e.g., Rhinolophids) have high  
655 numbers of as yet undescribed cryptic species (Mayer et al., 2007). The lack of  
656 distribution data may be explicitly linked to the lack of funding in most biodiverse countries  
657 (McClanahan and Rankin, 2016), hindering effective assessment in countries most in  
658 need. Conversely within Europe, the UNEP-EUROBATS and EU habitats directive  
659 provides guidance and regulations, which comprehensively include the protection and  
660 monitoring of bats and caves in its territory and member states making parallel and  
661 equitable policies for large scale protection (European Commission, 2021;  
662 UNEP/EUROBATS, 2020). Policies and targets that accurately account for and include  
663 monitoring in threatened systems such as caves and karsts are urgently needed and

664 highlight a need for ecosystem-based conservation targets, as species-specific targets  
665 risk missing key habitats for neglected taxa (Hughes et al., 2021).

## 666 **Synthesis**

667 Finally, to protect and conserve these ecosystems it is crucial that we identify  
668 priorities in species and habitat-level, and map vulnerable cave habitats with the highest  
669 biodiversity and distinctiveness. We illustrate a comprehensive index to integrate facets  
670 of diversity and risk to provide a simple and scalable approach to prioritising caves for  
671 protection and delineating between those in need of urgent intervention (high diversity but  
672 high threat) and those which whilst not yet threatened require monitoring to ensure they  
673 remain protected (Figure 7). Whilst further data is needed, especially for data-poor,  
674 species-rich regions, relying on IUCN data alone risks misleading effective priorities (e.g.,  
675 spatial mismatches in the Indomalayan region) (Martín-López et al., 2011; Milner-Gulland  
676 et al., 2006). The IUCN Red list must be utilised alongside other tools and measures of  
677 decision making; thus, we advocate an integrative conservation approach that  
678 synthesises various dimensions of conservation. Furthermore, habitat focused indices  
679 complement other recent initiatives such as the IUCN Red List of ecosystems, and new  
680 IUCN ecosystem typology, but also includes high-resolution data which can be  
681 challenging to include in broader scale indices.

682 Cave ecosystems host both high diversity and site-specific endemism, and are  
683 used by up to 30% of all bat species, yet are rarely included in global priorities (Mammola  
684 et al., 2019a; Sánchez-Fernández et al., 2021). The persistence of high levels of  
685 biodiversity in caves is linked to more pristine cave environments with less anthropogenic  
686 pressure. Large spatial coverage (e.g., biome-dependent) analysis decreases the

687 evenness in priority distribution in comparison to site-level, thus its conservation  
688 application to protect caves is scale dependant (e.g., national-level or biome-wide  
689 protection), and scalability in indices is also important. Conservation decision making  
690 depends on the clear delineation between what is important and urgent to develop  
691 priorities, as funding and resources are limited and often focus on a subset of taxa which  
692 may not be representative (Gordon et al., 2019; Joseph et al., 2009; Wilson et al., 2007).  
693 Ideally, priority-setting should encompass not only the aspects of biological diversity and  
694 threats but should also include the country's socio-economic capacity, the implementation  
695 cost, and political attributes. Thus, it is imperative to set realistic and cost-effective  
696 priorities based on areas with higher risks, vulnerabilities and could protect larger  
697 communities (Joseph et al., 2009; Rudd et al., 2011).

698         Our study identifies gaps and priorities for bat cave conservation. We highlight just  
699 how many high diversity caves are currently threatened (red caves) and those that are  
700 currently at low risk but biotically important (green caves), which reflects what form of  
701 intervention may be needed for different sites. To address global gaps for effective  
702 conservation and develop more rigorous priorities, collaborative efforts among bat  
703 scientists are necessary to enable the development of an effective international agenda  
704 and nationally adaptable goals for cave prioritisation and assessment. This framework  
705 calls for standardisation of methods of bat cave assessment and monitoring based on key  
706 indicators for diversity and threat, and for this data to be more widely shared to facilitate  
707 better conservation and policy. Furthermore, maintaining cave bat diversity relies on their  
708 inclusion into conservation agendas and priorities, and the use of science-based targets  
709 and frameworks, and synthesising conservation effectiveness (e.g., Conservation

710 Evidence Initiative (Conservation Evidence, 2021)) to ensure that approaches such as  
711 that detailed here can be effectively applied to enable key sites to be identified, and  
712 appropriately protected on a global basis, even despite a lack of data in many regions.  
713 Ultimately, developing effective decisions requires comparable data and standardised  
714 frameworks to enable its translation into policy and practice and effectively protect  
715 threatened species and facilitate sustainable development.

#### 716 **Data accessibility statement**

717 The data which are used in this study have been provided in the supplementary material  
718 accompanied with this manuscript.

#### 719 **Declaration of competing interest**

720 The authors declare that they have no known competing financial interests or personal  
721 relationships that could have appeared to influence the work reported in this paper.

#### 722 **CRedit authorship contribution statement**

723 KCT and ACH contributed equally to the work. KCT and ACH conceptualised and  
724 designed the study. KCT, HFMO, and ACH collected the data. KCT and ACH analysed  
725 the data. KCT and ACH led the manuscript writing. KCT performed data visualisation.  
726 ACH supervised the project. All authors contributed to the editing, reviewing, and  
727 approved the final version of the manuscript.

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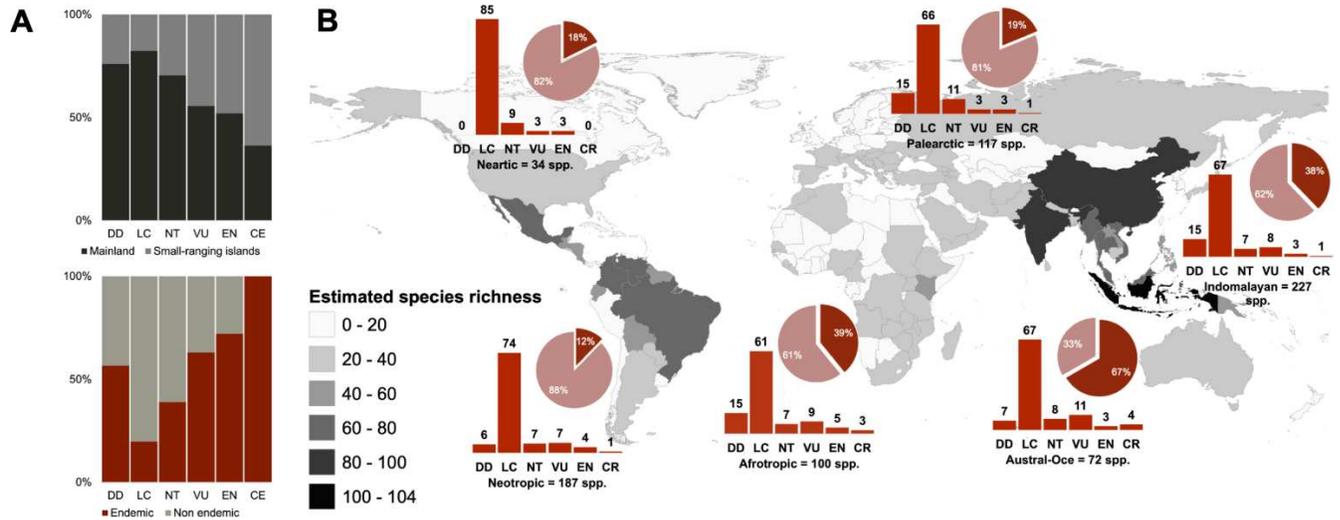
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1216 **List of figures**

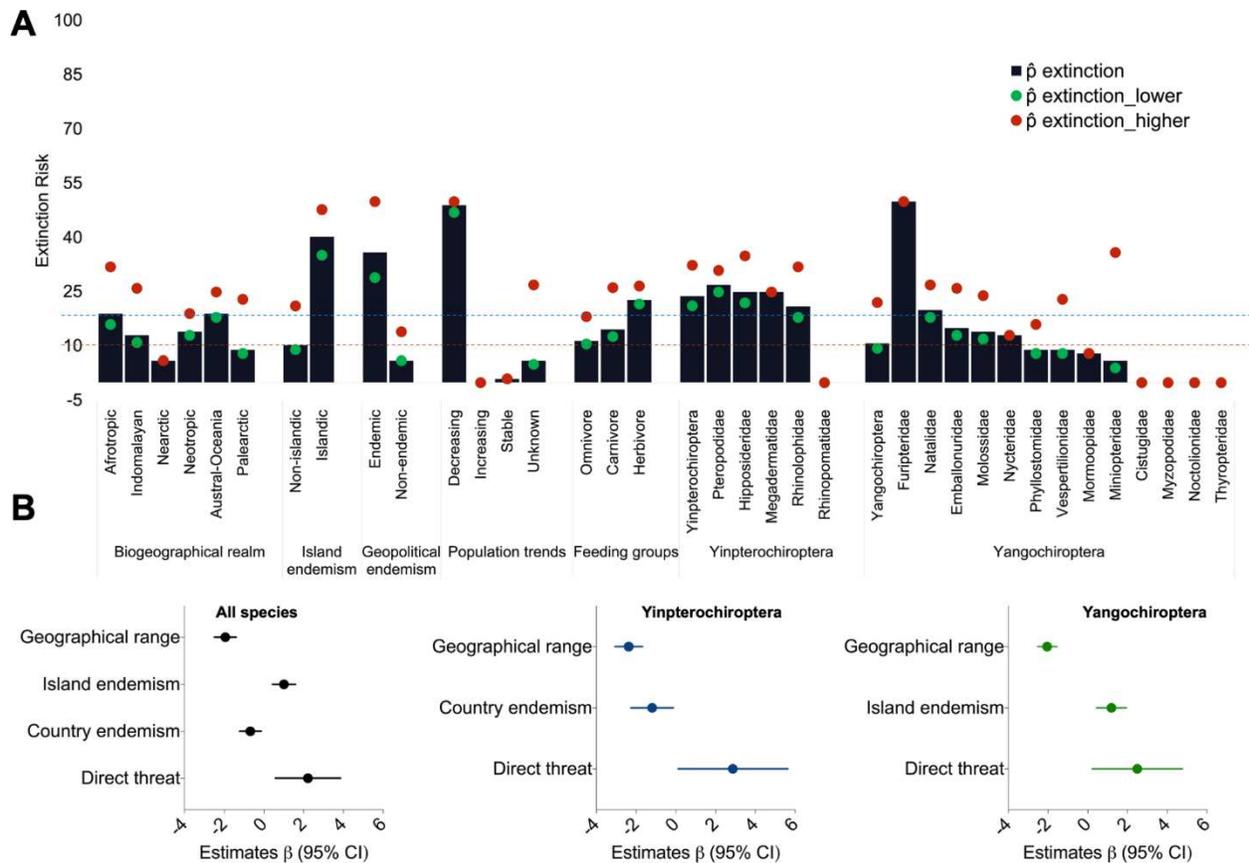


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1218 **Figure 1. Species diversity and distribution of cave-dwelling bats.** IUCN-based  
 1219 conservation status according to (a) country and island endemicism, and (b) taxonomic  
 1220 distribution of estimated species richness of cave bats per country, and proportion of  
 1221 endemic species (in pie graph) and threatened (in the bar graph) of compared by  
 1222 biogeographical realm. The proportion of country endemic species differed regionally ( $\chi^2$   
 1223 = 93.49,  $df = 5$ ,  $P < 0.0001$ ) and is highest in the Indomalayan region (38%,  $N = 86$   
 1224 spp.). The highest number of nationally endemic species was recorded in Madagascar  
 1225 ( $N = 23$  spp., % endemism = 82%), Indonesia ( $N = 21$  spp., % endemism = 20%), and  
 1226 Australia ( $N = 18$  spp., % endemism = 55%) (See also Data S2).

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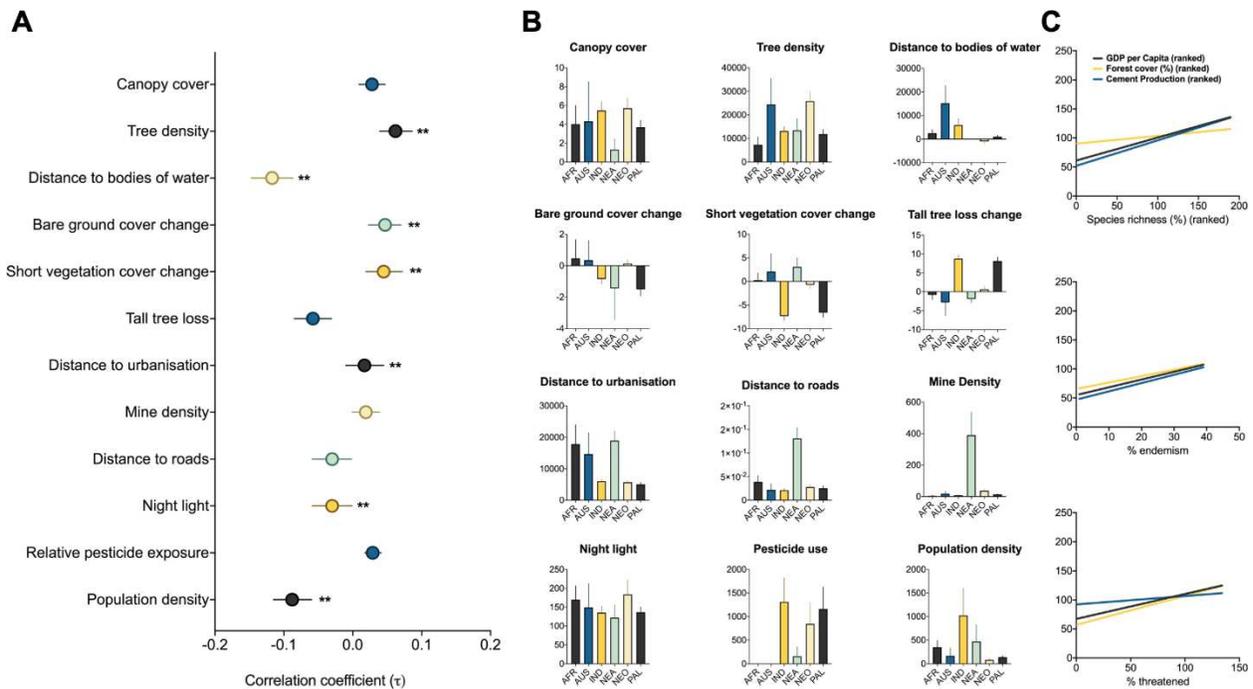
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1230 **Figure 2. IUCN-based extinction risk.** (a) Estimated extinction risks ( $\hat{\beta}_{extinction}$ ) of  
 1231 global cave-dwelling bat species compared by biogeographical realm, island endemism,  
 1232 geopolitical endemism, population trends, trophic groups, sub-orders and families.  
 1233 Estimated extinction proportion ( $\hat{\beta}_{extinction}$ ) for global species (blue dashed line) and all  
 1234 cave-dwelling bats (red dashed line) is provided. All computed values are supplemented  
 1235 in Table S2. (B) Estimate coefficients of significant determinants extinction risks of all  
 1236 global species, and amongst suborders Yinpterochiroptera and Yangochiroptera.  
 1237 Summary of binomial generalised linear mixed (GLMs) explaining species extinction  
 1238 risks is provided in Table S3.

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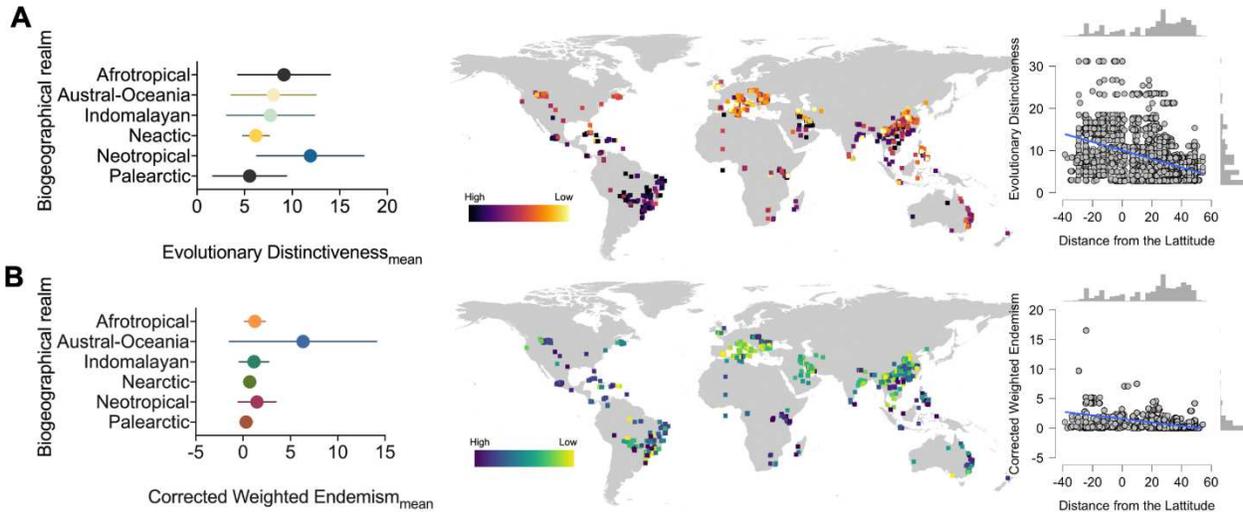
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1243 **Figure 3. Relationship, extent, and distribution of threats amongst assessed**  
 1244 **global cave sites.** (a) Correlation coefficients ( $\tau$  and 95% CI intervals) between  
 1245 landscape features and threats and species cave biotic scores, (b) biogeographic  
 1246 comparison of mean intensity of vulnerabilities to landscape features and threats. When  
 1247 compared regionally, the mean intensity of landscape features and threat variables  
 1248 differed significantly except for relative exposure to pesticide. Bat caves in the  
 1249 Afrotropical, Austral-Oceania and Nearctic regions showed the highest vulnerability to  
 1250 distance to urban areas. While Nearctic caves showed the nearest distance to roads  
 1251 and the highest mapped mine density. The mean proportion of bare ground cover  
 1252 change is low in arid regions of the Afrotropical and Austral-Oceania. Whereas  
 1253 population density and relative pesticide exposure are exceedingly high in Indomalayan  
 1254 caves, highlighting potentially high deforestation and loss of natural habitats. (c) Global  
 1255 concordance of between country socioecological indicators (country GDP, Forest cover,  
 1256 cement export production) and species attributes (up to down: country species richness,  
 1257 endemism, and proportion of threatened species). (values with \*\* showed significance  
 1258 at  $P < 0.05$ ).



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1260 **Figure 4.** Comparison of bat cave evolutionary distinctiveness (A) and endemism (B)  
 1261 across biogeographical realm and the latitudinal relationship of ED and CWE (right).

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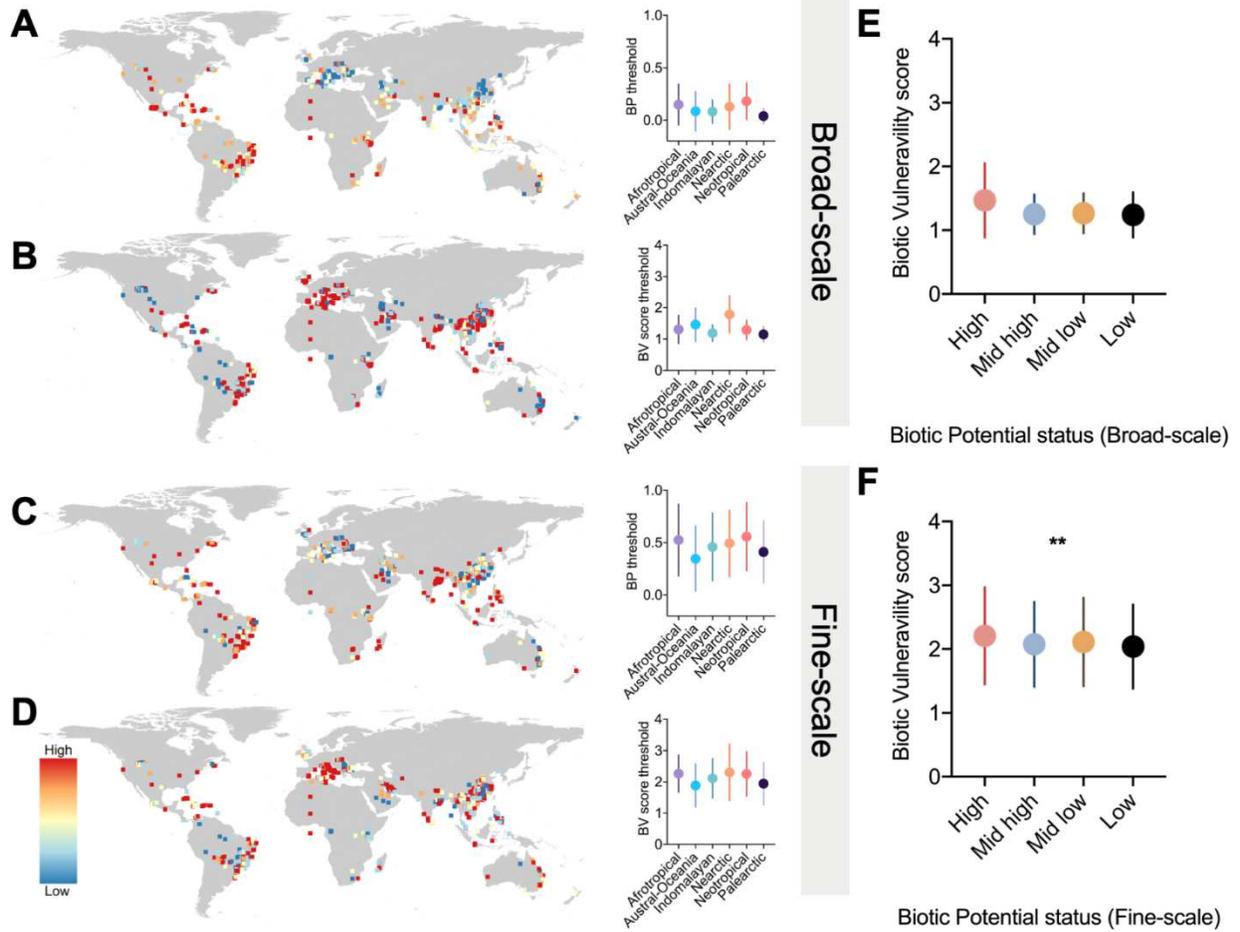
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1271 **Figure 5. Distribution and relationships of cave biotic potential and vulnerability**  
 1272 **worldwide.** Broad-scale (A) Biotic Potential (BP), (B) Biotic Vulnerability (BV); and Fine-  
 1273 scale (C) Biotic Potential (BP), and (D) Biotic Vulnerability (BV) (All values significantly  
 1274 differed; error bar represents mean  $\pm$  SD). (E-F) relationship of cave Biotic Potential  
 1275 status to Biotic Vulnerability scores compared in both broad-scale and fine-scale  
 1276 analyses (values with \*\* showed significance at  $P < 0.05$ ; error bars represent mean  $\pm$   
 1277 SD).

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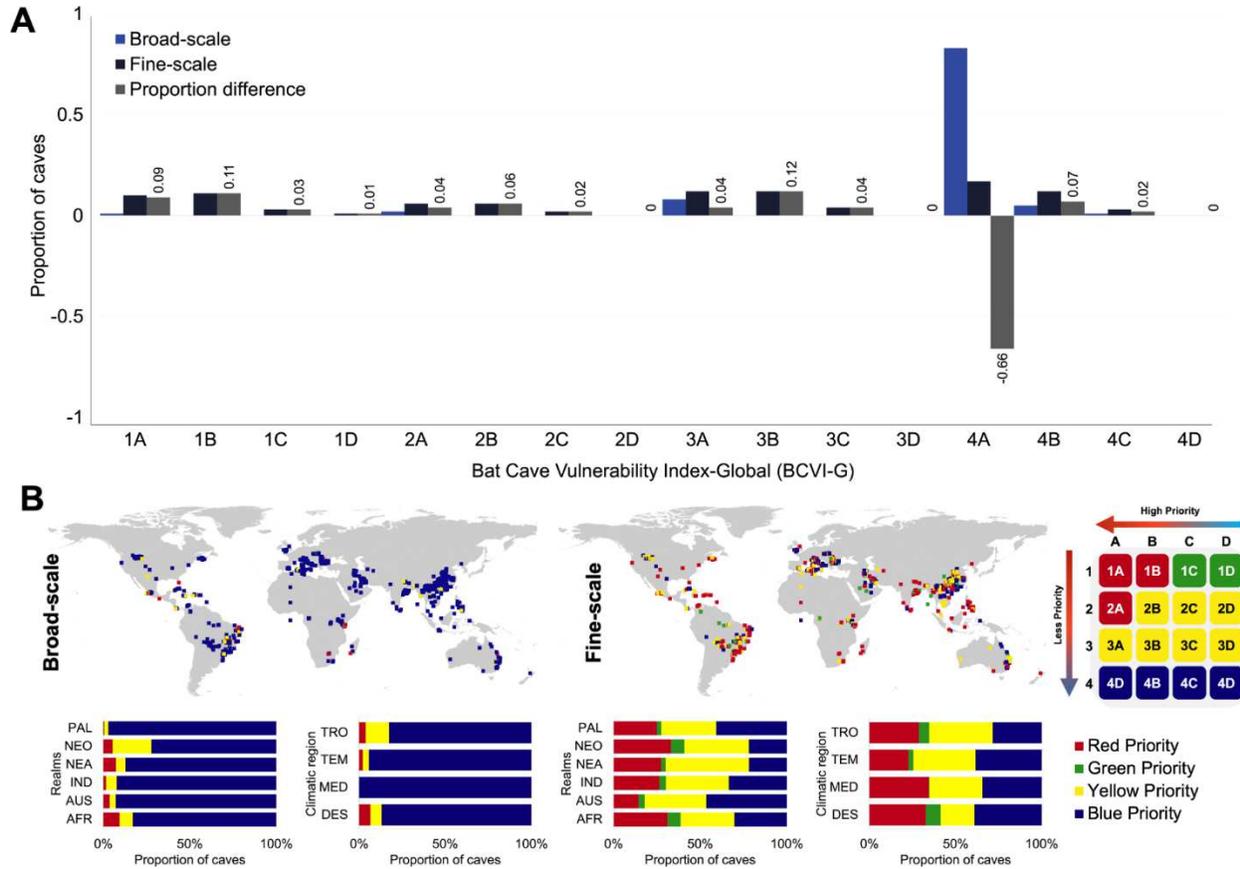
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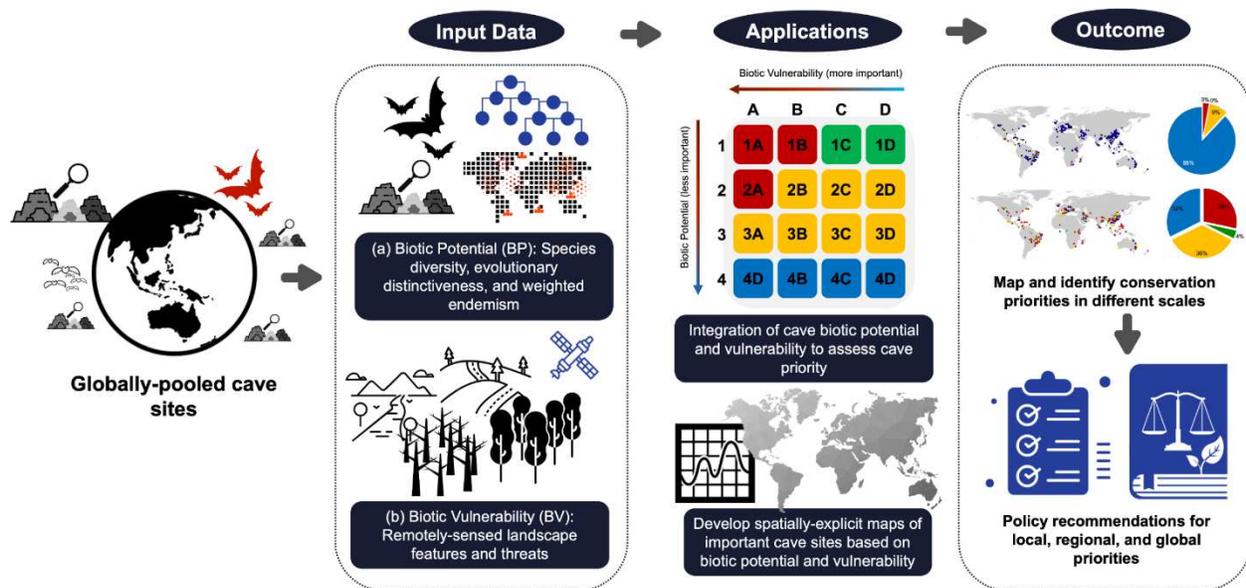
1286 **Figure 6. Global Bat Cave Vulnerability Index and Priorities.** (a) Proportion and  
 1287 proportion differences of cave Bat Cave Vulnerability Index- Global (BCVI-G) analysed  
 1288 in broad-scale and fine-scale. The number and percentage of caves according to BCVI-G  
 1289 scales is supplemented in Data S3. Spatial conservation priorities in (b) broad-scale  
 1290 and fine-scale based on Bat Cave Vulnerability Index- Global (BCVI-G). Broad-scale  
 1291 shows poor performance in showing high priority caves compared to fine-scale  
 1292 prioritisation. Proportions are compared across biogeographical realms and climatic  
 1293 regions. Summaries BCVI-G based priorities are being supplemented in Table S6.

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1299 **Figure 7. Framework for prioritising bat caves for conservation.** The schematic  
 1300 diagram showing the frame of the Bat Cave Vulnerability Index-Global (BCVI) in  
 1301 identifying and setting priorities for bat caves for conservation on a global scale. The  
 1302 process starts with the pooling of caves sites for assessment assess cave priorities. The  
 1303 input data includes the bat cave diversity (BP) and remotely senses landscape features  
 1304 to assay threats (BV). The index is categorised in four (4) levels with red caves are high  
 1305 priority sites containing the highest diversity and threats and requires immediate  
 1306 conservation actions, while blue caves are the lowest priority sites with the least  
 1307 biodiversity and high to low threat, but no immediate conservation attention is needed.

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1316 **List of Supplementary Information**

1317 **Data S1.** Complete list of cave-dwelling bat species pooled from the International Union  
1318 for the Conservation of Nature Red List (IUCN Red List) version 2020. The supplementary  
1319 dataset contains complete taxonomic information, habitat preference, biogeographical  
1320 range, species distribution, conservation status, endemism, ecological traits, and  
1321 threatening process.

1322 **Data S2** Country-level species diversity of cave-dwelling bats including estimated species  
1323 richness, endemism, and proportion of threatened species. The supplementary data also  
1324 contains the country-level socioecological status including the rate of cement production,  
1325 forest cover (%), and country Gross Domestic Production (GDP).

1326 **Data S3.** Complete list of cave sites and species diversity analysed for prioritisation using  
1327 Bat Cave Vulnerability Index-Global (BCVI-G). The results of BCVI-G prioritisation in  
1328 broad-scale and fine-scale are also provided.

1329 **Figure S1.** Supplementary Figure 2. Distribution of global dominant threats (A) and  
1330 biogeographical patterns (frequency) of the threatening process (B) and mean threat  
1331 index (C). Complete threat distribution per species is supplemented in Supplementary  
1332 Data 1.

1333 **Figure S2.** Effects of geographic range, adult body mass, trophic group and island  
1334 endemism on species risks to key threatening process. Complete results of the models  
1335 are supplemented in Supplementary Table 4.

1336 **Table S1.** Distribution of conservation status across families.

1337 **Table S2.** Extinction risks of cave-dwelling bats across different taxonomic and ecological  
1338 dimensions

1339 **Table S3.** Summary of Binomial Generalised Linear Mixed (GLMs) explaining species  
1340 extinction risk of (A) global species, and in between suborders (B) Yinpterochiroptera,  
1341 and (C) Yangochiroptera.

1342 **Table S4.** Summary of threatening process across the biogeographical realm and  
1343 Species Threat Index (SPI) in each ecological dimension.

1344 **Table S5.** Summary statistics of Generalised Linear Models (GLM) predicting the  
1345 independent effects of species traits: geographical range, adult body mass, and trophic  
1346 levels on the vulnerability of species to top threatening process in a global scale.

1347 **Table S6.** Distribution (cave count) of BCVI in broad-scale and fine-scale across  
1348 Biogeographical realm and climatic regions.

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## Supplementary Files

This is a list of supplementary files associated with this preprint. Click to download.

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- [DataS2.xlsx](#)
- [DataS3.xlsx](#)