

A standard protocol for harvesting biodiversity data from Facebook

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37

38 **Author contributions**

39 SC conceptualised the idea; SC developed the method; SC, SuA, ShA, PD, MML, MR, and AS
40 collected the data; SC did the analysis; everyone contributed to the analysis; SC wrote the
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46 **Data availability**

47 Our extracted Facebook data are publicly available (Chowdhury et al., 2022b).

48

49 **Abstract**

50 1. The expanding use of citizen science platforms has led to an exponential increase in
51 biodiversity data in global repositories. Yet, our understanding of species distribution
52 remains patchy for most of the world. Social media data has the potential to reduce the
53 global biodiversity knowledge gap. However, practical guidelines and standardised pipelines
54 to harvest such data sources are still missing.

55 2. Here, we provide a standardised framework to extract species distribution records from
56 Facebook groups that allow access to their data following data privacy and protection
57 safeguards. Some countries actively use and moderate Facebook groups to share species
58 records. We present how to structure keywords, search for species photographs, and
59 georeference localities for such records. We further highlight some challenges users might
60 face when extracting species distribution data from Facebook and suggest potential
61 solutions.

62 3. Following our proposed framework, we present a case study on Bangladesh's biodiversity
63 – a tropical megadiverse South Asian country. We scraped nearly 45,000 unique locality
64 data for 967 species, with a median of 27 records per species. About 12% of the distribution
65 data were for threatened species, which represent 27% of all species. We also obtained data
66 for 56 Data Deficient species.

67 4. If carefully harvested, social media data can significantly reduce global biodiversity
68 knowledge gaps. Consequently, developing an automated tool to extract and interpret
69 social media biodiversity data is an essential research priority.

70

71 **Introduction**

72 Amid the sixth mass extinction, many species worldwide are dramatically declining – 28% of
73 all assessed species in the IUCN Red List are threatened with extinction (Dirzo et al., 2014;
74 Pimm et al., 2014; Murali et al., 2023). A recent study that analysed population trends for all
75 IUCN Red List assessments revealed that 48% of species are declining, 49% are in stable
76 condition, and only 3% of assessed species are increasing (Finn et al., 2023). The Living
77 Planet Index report reveals an average 69% decrease in monitored wildlife populations since
78 1970 (WWF, 2022). However, global assessments are highly biased towards certain taxa or
79 regions (Miqueleiz et al., 2020). For example, 85% of described reptile species have been
80 assessed by the IUCN Red List (Meiri et al., 2023) vs ca. 1% of arthropod species (IUCN,
81 2023). This severe discrepancy reflects a long-known bias in research interests (Clark & May,
82 2002; Cardoso et al., 2011; Di Marco et al., 2017) and can be partially attributed to missing
83 and inadequate distribution data that is fundamental for species threat assessments (Beck
84 et al., 2014; Hughes et al., 2021). For example, a taxonomic revision and distribution sample
85 of beetle species found that 53% of the 186 species were only known from a single locality,
86 and 13% from a single specimen (Stork, 1997). These sorts of biases are known as Linnean
87 shortfalls (taxonomic knowledge gaps) and Wallacean shortfalls (distribution knowledge
88 gaps; Hortal et al., 2015; Diniz-Filho et al., 2023). Such poor representation of species
89 records is also prominent in the most extensive biodiversity repository - Global Biodiversity
90 Information Facility (GBIF) – which contains locality data only for 10% of described insect
91 species (Chowdhury et al., 2023a).

92 To bridge this knowledge gap, many initiatives are using knowledge from the general public
93 (Hochkirch et al., 2015). These data collection initiatives are commonly known as citizen
94 science or community science, in which people share their species observation records
95 through different, often online, applications (Conrad & Hilchey, 2011; Bela et al., 2016;
96 Chandler et al., 2017; Pocock et al., 2019; Callaghan et al., 2021). These observations, in
97 many cases, are eventually deposited in large online repositories (e.g. GBIF; Roy et al.,
98 2018; Callaghan et al., 2022). There are hundreds of such citizen science applications
99 globally that have greatly improved our understanding of biodiversity patterns in recent
100 years. Since 2007, there has been a 12-fold increase in biodiversity data in GBIF (Heberling
101 et al., 2021; GBIF contains > 2.3 billion species occurrence records as of 9 June 2023).
102 However, despite increases in the volume of biodiversity data available, geographic and
103 taxonomic knowledge gaps on species distributions remain (Hughes et al., 2021). Most
104 biodiversity observation records are from Europe and North America, resulting in major
105 sampling and observation biases (Ramírez et al., 2022). While most species are found in
106 tropical forests, our knowledge of the biodiversity of these regions is extremely limited
107 (Collen et al., 2008; Hortal et al., 2015; Meyer et al., 2015; Hughes et al., 2021; Chowdhury
108 et al., 2023b). Though it comprises less than 2% of the Earth's area, nearly 36% of the
109 biodiversity data in GBIF come from the United States. Conversely, Brazil, which has a
110 similar land area to the USA and is the most biodiverse country on Earth, is represented by
111 only 0.8% of the records in GBIF (accessed on 22 May 2023). Therefore, new approaches
112 and methods are needed to overcome the Linnean and Wallacean shortfalls.

113 With the increasing popularity of social media and the growing availability of digital phones
114 and fast internet, many people post biodiversity observations on different social media
115 platforms (e.g., Facebook, Twitter; Di Minin et al., 2015; Andrachuk et al., 2019; Toivonen et
116 al., 2019) that don't necessarily make it to the existing pipelines to GBIF. Among these
117 platforms, Facebook has become the most popular social media network (Anderson et al.,
118 2012). There are thousands of biodiversity observation groups on Facebook globally, with a
119 wealth of species distribution information, often with more in-depth data than available in
120 global biodiversity repositories (Chowdhury et al., 2023b). For example, by scraping a single
121 Facebook group for the butterflies of Bangladesh, Chowdhury et al. (2021) obtained about
122 35 times more distribution records from Facebook than were deposited in GBIF at the time.
123 Moreover, Facebook also contains data regarding many unique species that are absent from
124 GBIF altogether (Chowdhury et al., 2023b). Such data have also been shown to be key for
125 improved spatial conservation prioritisation (Chowdhury et al., 2023c). The utility of
126 Facebook as a biodiversity repository is possible due to volunteer contributions of
127 moderators and administrators of Facebook groups that help users identify species
128 (Chowdhury et al., 2023b; Marceno et al., 2021). Moreover, many scientists are unaware of
129 the great biodiversity potential found in social media data or are unfamiliar with pipelines to
130 extract such data (Chowdhury et al., 2023c).

131 Several studies explored the importance of using Facebook data to fill the global biodiversity
132 data shortfall (e.g., Chamberlain, 2018; Marcenò et al., 2021; Chowdhury et al., 2023b;
133 O'Neill et al., 2023). While having a standard data collection protocol is essential to improve
134 conservation assessments, there is no such protocol that researchers can follow to collate
135 species distribution data from Facebook. Here, we provide a complete pipeline to harvest
136 quality biodiversity data from Facebook groups, which were accessed by receiving
137 permission from the groups' administrators, for Bangladesh as a case study. We discuss i)

138 the keyword formulation process, ii) how to search inside Facebook groups and extract and
139 filter species photographs, and iii) how to georeference the location information. Finally, we
140 point out the potential of species distribution data obtained from Facebook and some
141 challenges that users might face when extracting species distribution data from Facebook.

142

143 **Methods**

144 *Data extraction framework*

145 Extracting species locality information from Facebook contains three steps: group selection,
146 data extraction, and georeferencing the record location (Figure 1a).

147 *Step 1 - Group selection*

148 Before starting the data extraction process, it is important to filter the relevant groups using
149 a systematic search. Search keywords could contain a combination of taxon and country
150 names. For example, if there is an interest in extracting data for 'birds' from 'Bangladesh',
151 the keywords can be: 'Bird Bangladesh'. Here, singular/plural form or capitalisation of words
152 do not make any substantial difference - except in their order of appearance. We
153 recommend using different combinations of keywords (including region names) when listing
154 relevant Facebook groups. After formulating keywords, the search can be conducted using
155 the Facebook search function to identify relevant groups (Figure 1).

156 *Step 2 - Data extraction*

157 When extracting data, the authors should carefully consider the privacy of the group:
158 whether it is public, private or secret and maintain the data usage policy (Di Minin et al.,
159 2021). Authors should also prepare a complete list of species that need manual search.

160 When a relevant group is selected, the next step is to extract species information. This can
161 be performed by searching for species locality records inside individual groups. Searching
162 within Facebook groups is similar to searching in engines like Google Scholar: species names
163 can be searched with the search function of the selected Facebook group ('magnifying glass'
164 icon inside the Facebook group), which includes options to restrict the search (e.g., year;
165 Figure 1).

166 During data extraction (from posts, but not photographs), some issues may be encountered.
167 For example: i) species misidentification in the caption, ii) inconsistencies with species
168 names (some users use the scientific name, while others use the English or local name), iii)
169 Facebook becomes too slow when the keyword searches produce too many search results,
170 and iv) search results also include erroneous species due to partial keyword overlap (e.g.,
171 butterfly named 'common pierrot' could appear when searching for 'common jay'). To
172 handle misidentification issues, we recommended double-checking species identifications
173 before extracting data. In active Facebook groups, moderators, administrators, or other
174 users often verify individual photographs and provide suggestions when required. This is
175 particularly relevant when they help users identify photographs and confirm locality
176 information. Such information can often be found in the comment section of each
177 photograph. To control naming inconsistencies, we recommend searching by scientific
178 name, English name, and local names for each species in each group. Another issue is that
179 the same photographs might appear in different searches due to the partial keyword

180 overlap. To control this issue, researchers could use a unique identifier for the users and
181 then remove duplicate information. To expedite the search process during an individual
182 search, we recommend restricting the search by year using the 'Date Posted' tab on the left-
183 hand side of the screen. Here, individual years (e.g., 2021) can be filtered for each search,
184 and then the process should be repeated until there are no further search results. Finally,
185 duplicate records can be removed.

186 *Step 3 - Georeferencing*

187 Facebook lacks an automated georeferencing system. Using the location information from
188 species photographs (Figure 1), any mapping software (e.g., Google Map, Google Earth,
189 ArcGIS) can be used to extract the latitude and longitude information. Conveniently, Google
190 Place ID API ([https://developers.google.com/maps/documentation/places/web-](https://developers.google.com/maps/documentation/places/web-service/place-id)
191 [service/place-id](https://developers.google.com/maps/documentation/places/web-service/place-id)) enables automatic georeferencing for many locations. It can also be done
192 using the 'geocode' function of the 'ggmap' R package (Kahle and Wickham, 2013).

193

194 *Case study on Bangladesh*

195 *Species list*

196 To illustrate the application of our proposed framework, we performed a case study for
197 extracting biodiversity data from Bangladesh's dedicated Facebook groups. We compiled a
198 checklist of the animals of Bangladesh with a known conservation status (1,619 species)
199 from the most recent national Red List (IUCN Bangladesh, 2015). While this is the most
200 comprehensive checklist for the biodiversity of Bangladesh, the list is already eight years
201 old, and several other undocumented species remain. From the Red List database, we
202 extracted group names (e.g., Birds, Butterflies), Order, Family, scientific name, English
203 name, local name, and the Red List assessment status in Bangladesh. While cross-checking
204 the data, we noticed that the Family information for butterflies was old and did not match
205 the Family names found in GBIF. We updated family names following the GBIF taxonomy to
206 handle this issue. Specifically, we updated the information for four Families (Acraeidae,
207 Amathusiidae, Danaidae, and Satyridae) and moved them under the Family Nymphalidae.
208 Finally, we removed regionally extinct species (assessed as RE in the national Red List) from
209 the species list.

210 *Step 1 - Group selection*

211 First, we searched for each taxonomic group (e.g., Bird) and added the country name
212 (Bangladesh) at the end (e.g., Bird Bangladesh). From our previous experiences, we were
213 aware that some Facebook groups contain the term 'biodiversity'. Consequently, we also
214 searched Facebook using the keywords 'biodiversity Bangladesh'. Altogether our search
215 keywords included the following seven combinations: 'Amphibian Bangladesh', 'Bird
216 Bangladesh', 'Butterfly Bangladesh', 'Crustacean Bangladesh', 'Fish Bangladesh', 'Mammal
217 Bangladesh', and 'Reptile Bangladesh'. Based on the search results, we filtered the most
218 popular Facebook groups, for each taxon, based on i) moderation activity (e.g., whether
219 group moderators help users with species identification), ii) group rules (if the group has
220 strict rules about the location and date of the photographs), and iii) group activities (if
221 members post every day).

222 *Step 2 - Data extraction*

223 We followed a range of approaches when extracting species distribution data from
224 Facebook groups. Before starting the data extraction process, we conducted a test search
225 with some common and rare species. To control naming inconsistencies, we searched each
226 species with its scientific name, English name, and local name in Bengali, which led to three
227 rounds of search for each species in each group. To expedite the search process, we
228 restricted searches by year with the 'Date Posted' tab. Here, we filtered individual years
229 (e.g., 2021) for each search and extracted all relevant results for that year. We repeated this
230 process until there were no more new search results. Finally, we removed duplicate records
231 from the compiled datasheet (see above).

232 From each post, we extracted the following information: species name (search keyword),
233 life stage (e.g., adult, egg), date (day/month/year), location information, and the name of
234 the photographer. For quality control, we double-checked if the species identification was
235 correct before extracting its information. We also skimmed through the comments of
236 individual photographs. We excluded photographs if i) the photograph was not from
237 Bangladesh, ii) the photograph was unclear, iii) the species was not identified up to the
238 species level, iv) the locality information was missing, and v) if the area of the location was
239 over 100 km².

240 *Step 3 - Georeferencing*

241 As stated above, Facebook lacks an automated georeferencing system. Using the location
242 information from each post, we searched it in Google Map
243 (<https://www.google.com/maps>), selected a random point within that area, and extracted
244 the latitude and longitude (in decimals). Considering that we discarded photographs if the
245 specified location was an area over 100 km², the precision uncertainty of these extracted
246 distribution records was within 10 km.

247

248 **Results**

249 **Data harvested from Bangladesh**

250 Overall, we identified 42 Facebook groups for different animals in Bangladesh
251 (supplementary Table S1). While 15 of these groups are on birds, only one is on odonatans
252 (Figure 2). There are only four groups with > 50,000 members. Two groups are focused on
253 multiple taxa, one on birds and one on reptiles (Figure 2; supplementary Table S1).

254 Of the 42 Facebook groups, we chose the seven most popular Facebook groups: Birds
255 Bangladesh (<https://www.facebook.com/groups/2403154788>); Deep Ecology And Snake
256 Rescue Foundation (<https://www.facebook.com/groups/959896627527624>); Biodiversity of
257 Bangladesh; (<https://www.facebook.com/groups/249240636186853>); Butterfly Bangladesh;
258 (<https://www.facebook.com/groups/488719627817749>); Mammals of Bangladesh;
259 (<https://www.facebook.com/groups/647662968655338>); Amphibians and Reptiles of
260 Bangladesh; (<https://www.facebook.com/groups/560709511527645>); Biodiversity of
261 Greater Kushtia (<https://www.facebook.com/groups/244807066739477>).

262 We collated 44,726 occurrence records for 967 species, ranging between 1-719 records per
263 species. These data included 45 amphibian species, 494 bird species, 265 butterfly species,

264 72 mammal species, and 91 reptile species (Figures 3A, 4). We could not locate any species
265 occurrence records for fishes or crustaceans.

266 While the median occurrence records per species were 27, it varied substantially amongst
267 taxa: the median species occurrence records were 54 for birds, 33 for butterflies, 5 for
268 mammals, and 3 for reptiles. There were 196 species with 80 or more records, of which only
269 seven were butterflies (no threatened species), and the rest were birds (one threatened
270 species, *Threskiornis melanocephalus*, Vulnerable). The following were the most popular
271 species from each group: *Alcedo atthis* (birds, 719 records), *Danaus chrysippus* (butterflies,
272 107 records), *Prionailurus viverrinus* (mammals, 66 records), *Xenochrophis piscator* (reptiles,
273 39 records), *Duttaphrynus melanostictus* and *Polypedates leucomystax* (amphibians, 14
274 records).

275 The number of occurrence records grew substantially with time, with some random
276 fluctuations (Figure 3B). Although Facebook started in the early 2000s, we obtained many
277 records before that date. For birds, there were records available from 1978. For other taxa,
278 the records started in 1992 for reptiles, 1998 for mammals, 2004 for butterflies and 2005 for
279 amphibians. For all taxa, the number of species occurrence records increased markedly from
280 the start date to early 2020; however, the numbers rapidly declined afterwards, possibly
281 due to the COVID-19 pandemic (Figure 3B).

282 We obtained distribution data for 260 threatened species (27% of species), 651 non-
283 threatened species (67% of species), and 56 Data Deficient species (6% of species) (see the
284 supplementary Table S2). While 12% of the species distribution data was for threatened
285 species, it varied substantially across taxa. For mammals, 70% of the occurrence records
286 were for threatened species, 47% for butterflies, 14% for reptiles, 6% for amphibians, and
287 3% for birds (Figure 4A). The pattern was different when considering the proportion of
288 species numbers within taxa recorded, both for threatened and non-threatened species. For
289 butterflies, we obtained records for twice more species that were classed as threatened
290 (167) than non-threatened (84). For other groups, threatened species contributed to 46% of
291 all mammal species, 20% for reptiles, 18% for amphibians, and 7% for birds (Figure 4B).

292

293 **Discussion**

294 The increasing number of citizen science applications are contributing to a sharp increase in
295 species distribution records (Heberling et al., 2021). However, despite this increase, a
296 substantial bias remains in our understanding of global biodiversity – the distribution of
297 tropical species remains overlooked (Hortal et al., 2015; Di Marco et al., 2017; Kühl et al.,
298 2020; Hughes et al., 2021). Here, we present a standardised protocol for Facebook data
299 extraction, and we demonstrate its use on Facebook biodiversity records from Bangladesh,
300 to assess how data obtained from this platform can help reduce the global bias in
301 biodiversity knowledge. We obtained nearly 45,000 records for 967 species from
302 Bangladesh, of which 27% are nationally threatened, and many are Data Deficient. We
303 found that over time, data increased sharply for all taxa; however, there was a substantial
304 decline in the amount of data in 2021 during the COVID-19 pandemic. We further provide
305 step-by-step guidelines to extract species locality data, which could be helpful for future
306 researchers to obtain local and global biodiversity data that could aid in conservation
307 assessments.

308 Facebook data has the potential to improve our biodiversity knowledge. However, there are
309 a few issues that users should consider (Di Minin et al., 2015; Toivonen et al., 2019;
310 Chowdhury et al., 2023b). First, extracting data from Facebook is a time-consuming and
311 multi-step process. On average, it took us nearly 33 minutes to complete the data extraction
312 process for a single species (i.e. ~532 hours in total). Second, Facebook photographs do not
313 contain precise or automatic geolocation functions, which results in coordinate
314 uncertainties. Third, maintaining the quality of posts in Facebook groups requires active
315 moderation activities, which include a high level of taxonomic expertise, and spending time
316 reviewing every photograph. While this is typical for Bangladesh, where administrators,
317 moderators, and users help maintain every photograph's quality, this might not be the case
318 for many countries. We present a standardised protocol to extract biodiversity data from
319 Facebook. While automatic extraction is not (yet) possible, developing such tools to
320 expedite the process would be useful. Likewise, developing tools that automatically deposit
321 species distribution records from Facebook to the global biodiversity repositories could be
322 instrumental in harvesting this source for biodiversity records (Jarić et al., 2020; Correia et
323 al., 2021). If more resources are allocated for this endeavour, people can be appointed to
324 check Facebook posts regularly, help administrators and moderators maintain group quality,
325 extract data regularly, and create a local database that will eventually be deposited into
326 global biodiversity repositories such as GBIF.

327 When using social media data for research purposes, inherent risks must be addressed to
328 protect individuals from potential harm, whether intentional or unintentional (Di Minin et
329 al., 2021). To mitigate these risks and ensure the users' safety, it is important to adopt
330 practices such as data minimisation, anonymisation, and strict data management protocols.
331 Employing risk-based approaches, such as conducting data privacy impact assessments, can
332 aid in identifying and minimising privacy risks for social media users. Besides, when sharing
333 the observation records, authors should carefully consider whether they should share
334 threatened species' locations (Sbragaglia et al., 2021). This not only showcases
335 accountability but also ensures compliance with data protection laws, to safeguard the
336 privacy of individuals involved in the research process (Di Minin et al., 2021).

337

338 **Conclusion**

339 While scientists worldwide are formulating innovative approaches to improve our
340 understanding of species distributions, there remain substantial taxonomic and regional
341 gaps and biases. Several studies have shown Facebook's importance in extracting
342 biodiversity records (e.g., Chowdhury et al., 2021, 2023b,c); however, there is no standard
343 pipeline that researchers can follow. Here, we provide detailed guidelines on preparing a list
344 of relevant Facebook groups and recommend approaches to search efficiently. While the
345 entire process still requires time, as we could not automate this yet, it is worth the effort,
346 given we obtained thousands of species occurrence records for a tropical biodiverse
347 country, Bangladesh. We recommend a concerted effort to create a global database
348 containing biodiversity groups on Facebook.

349

350 **References**

351 Andrachuk, M., Marschke, M., Hings, C., & Armitage, D. (2019). Smartphone technologies
352 supporting community-based environmental monitoring and implementation: a systematic
353 scoping review. *Biological Conservation*, 237, 430-442.

354 Beck, J., Böller, M., Erhardt, A., & Schwanghart, W. (2014). Spatial bias in the GBIF database
355 and its effect on modeling species' geographic distributions. *Ecological Informatics*, 19, 10-
356 15.

357 Bela, G., Peltola, T., Young, J. C., Balázs, B., Arpin, I., Pataki, G., ... & Bonn, A. (2016). Learning
358 and the transformative potential of citizen science. *Conservation Biology*, 30(5), 990-999.

359 Callaghan, C. T., Mesaglio, T., Ascher, J. S., Brooks, T. M., Cabras, A. A., Chandler, M., ... &
360 Young, A. N. (2022). The benefits of contributing to the citizen science platform iNaturalist
361 as an identifier. *PLoS biology*, 20(11), e3001843.

362 Callaghan, C. T., Poore, A. G., Mesaglio, T., Moles, A. T., Nakagawa, S., Roberts, C., ... &
363 Cornwell, W. K. (2021). Three frontiers for the future of biodiversity research using citizen
364 science data. *BioScience*, 71(1), 55-63.

365 Cardoso, P., Erwin, T. L., Borges, P. A., & New, T. R. (2011). The seven impediments in
366 invertebrate conservation and how to overcome them. *Biological Conservation*, 144, 2647-
367 2655.

368 Chamberlain, J. (2018). Using social media for biomonitoring: how Facebook, Twitter, Flickr
369 and other social networking platforms can provide large-scale biodiversity data. In *Advances*
370 *in Ecological Research* (Vol. 59, pp. 133-168). Academic Press.

371 Chandler, M., See, L., Copas, K., Bonde, A. M., López, B. C., Danielsen, F., ... & Turak, E.
372 (2017). Contribution of citizen science towards international biodiversity monitoring.
373 *Biological Conservation*, 213, 280-294.

374 Chowdhury, S., Zalucki, M. P., Hanson, J. O., Tiatragul, S., Green, D., Watson, J. E., & Fuller,
375 R. A. (2023a). Three-quarters of insect species are insufficiently represented by protected
376 areas. *One Earth*, 6(2), 139-146.

377 Chowdhury S, Aich U, Rokonuzzaman M, Alam S, Das P, Siddika A, ... & Callaghan CT.
378 (2023b). Increasing biodiversity knowledge through social media: a case study from tropical
379 Bangladesh. *BioScience*.

380 Chowdhury S, Fuller R, Ahmed S, Alam S, Callaghan C, Das P, ... & Bonn A. (2023c). Using
381 social media records to inform conservation planning. *Conservation Biology*. (in press)
382 <https://doi.org/10.1111/cobi.14161>.

383 Chowdhury, S., Jennions, M. D., Zalucki, M. P., Maron, M., Watson, J. E., & Fuller, R. A.
384 (2022a). Protected areas and the future of insect conservation. *Trends in Ecology &*
385 *Evolution*, 38, 85-95.

386 Chowdhury, S., Aich, U., Rokonuzzaman, M., Alam, S., Das, P., Siddika, A., ... & Fuller, R. A.
387 (2022b). Spatial occurrence data for the animals of Bangladesh derived from Facebook.
388 *Pangaea*. DOI: <https://doi.pangaea.de/10.1594/PANGAEA.948104>.

389 Chowdhury, S., Alam, S., Chowdhury, S. U., Rokonuzzaman, M., Shahriar, S. A., Shome, A. R.,
390 & Fuller, R. A. (2021). Butterflies are weakly protected in a mega-populated country,
391 Bangladesh. *Global Ecology and Conservation*, 26, e01484.

392 Collen, B., Ram, M., Zamin, T., & McRae, L. (2008). The tropical biodiversity data gap:
393 addressing disparity in global monitoring. *Tropical Conservation Science*, 1(2), 75-88.

394 Conrad, C. C., & Hilchey, K. G. (2011). A review of citizen science and community-based
395 environmental monitoring: issues and opportunities. *Environmental Monitoring and
396 Assessment*, 176, 273-291.

397 Correia, R. A., Ladle, R., Jarić, I., Malhado, A. C., Mittermeier, J. C., Roll, U., ... & Di Minin, E.
398 (2021). Digital data sources and methods for conservation culturomics. *Conservation
399 Biology*, 35(2), 398-411.

400 Di Marco, M., Chapman, S., Althor, G., Kearney, S., Besancon, C., Butt, N., ... & Watson, J. E.
401 M. (2017). Changing trends and persisting biases in three decades of conservation science.
402 *Global Ecology and Conservation*, 10, 32-42.

403 Di Minin, E., Fink, C., Hausmann, A., Kremer, J., & Kulkarni, R. (2021). How to address data
404 privacy concerns when using social media data in conservation science. *Conservation
405 Biology*, 35(2), 437-446.

406 Di Minin, E., Tenkanen, H., & Toivonen, T. (2015). Prospects and challenges for social media
407 data in conservation science. *Frontiers in Environmental Science*, 3, 63.

408 Diniz Filho, J. A. F., Jardim, L., Guedes, J. J., Meyer, L., Stropp, J., Frateles, L. E. F., ... & Hortal,
409 J. (2023). Macroecological links between the Linnean, Wallacean, and Darwinian shortfalls.
410 *Frontiers of Biogeography*. <https://doi.org/10.21425/F5FBG59566>.

411 Dirzo, R., Young, H. S., Galetti, M., Ceballos, G., Isaac, N. J., & Collen, B. (2014). Defaunation
412 in the Anthropocene. *Science*, 345(6195), 401-406.

413 Finn, C., Grattarola, F., & Pincheira-Donoso, D. (2023). More losers than winners:
414 investigating Anthropocene defaunation through the diversity of population
415 trends. *Biological Reviews*. <https://doi.org/10.1111/brv.12974>.

416 Heberling, J. M., Miller, J. T., Noesgaard, D., Weingart, S. B., & Schigel, D. (2021). Data
417 integration enables global biodiversity synthesis. *Proceedings of the National Academy of
418 Sciences*, 118(6), e2018093118.

419 Hochkirch, A., Samways, M. J., Gerlach, J., Böhm, M., Williams, P., Cardoso, P., ... & Dijkstra,
420 K. D. B. (2021). A strategy for the next decade to address data deficiency in neglected
421 biodiversity. *Conservation Biology*, 35(2), 502-509.

422 Hortal, J., de Bello, F., Diniz-Filho, J. A. F., Lewinsohn, T. M., Lobo, J. M., & Ladle, R. J. (2015).
423 Seven shortfalls that beset large-scale knowledge of biodiversity. *Annual Review of Ecology,
424 Evolution, and Systematics*, 46, 523-549.

425 Hughes, A. C., Orr, M. C., Ma, K., Costello, M. J., Waller, J., Provoost, P., ... & Qiao, H. (2021).
426 Sampling biases shape our view of the natural world. *Ecography*, 44(9), 1259-1269.

427 IUCN Bangladesh. 2015. Red list of Bangladesh: a brief on assessment result 2015, IUCN,
428 International Union for Conservation of Nature, Bangladesh Country Office, Dhaka,
429 Bangladesh.

430 Jarić, I., Correia, R. A., Brook, B. W., Buettel, J. C., Courchamp, F., Di Minin, E., ... & Roll, U.
431 (2020). iEcology: harnessing large online resources to generate ecological insights. *Trends in*
432 *Ecology & Evolution*, 35(7), 630-639.

433 Kahle, D. & Wickham, H. (2013). ggmap: Spatial Visualization with ggplot2. *The R Journal*,
434 5(1), 144-161.

435 Kühn, H.S., Bowler, D.E., Bösch, L., Bruelheide, H., Dauber, J., Eichenberg, D., ... & Bonn, A.
436 (2020) Effective biodiversity monitoring needs a culture of integration. *One Earth*, 3, 462-
437 474.

438 Marcenò, C., Padullés Cubino, J., Chytrý, M., Genduso, E., Salemi, D., La Rosa, A., ... &
439 Guarino, R. (2021). Facebook groups as citizen science tools for plant species
440 monitoring. *Journal of Applied Ecology*, 58(10), 2018-2028.

441 Meiri, S., Chapple, D. G., Tolley, K. A., Mitchell, N., Laniado, T., Cox, N., ... & Roll, U. (2023).
442 Done but not dusted: Reflections on the first global reptile assessment and priorities for the
443 second. *Biological Conservation*, 278, 109879.

444 Meyer, C., Kreft, H., Guralnick, R., & Jetz, W. (2015). Global priorities for an effective
445 information basis of biodiversity distributions. *Nature Communications*, 6(1), 1-8.

446 Miqueleiz, I., Bohm, M., Ariño, A. H., & Miranda, R. (2020). Assessment gaps and biases in
447 knowledge of conservation status of fishes. *Aquatic Conservation: Marine and Freshwater*
448 *Ecosystems*, 30(2), 225-236.

449 Murali, G., Iwamura, T., Meiri, S., & Roll, U. (2023). Future temperature extremes threaten
450 land vertebrates. *Nature*, 615(7952), 461-467.

451 O'Neill, D., Häkkinen, H., Neumann, J., Shaffrey, L., Cheffings, C., Norris, K., & Pettorelli, N.
452 (2023). Investigating the potential of social media and citizen science data to track changes
453 in species' distributions. *Ecology and Evolution*, 13(5), e10063.

454 Pimm, S. L., Jenkins, C. N., Abell, R., Brooks, T. M., Gittleman, J. L., Joppa, L. N., ... & Sexton,
455 J. O. (2014). The biodiversity of species and their rates of extinction, distribution, and
456 protection. *Science*, 344(6187), 1246752.

457 Pocock, M.J., Roy, H.E., August, T., Kuria, A., Barasa, F., Bett, J., Githiru, M., Kairo, J., Kimani,
458 J. & Kinuthia, W. (2019). Developing the global potential of citizen science: Assessing
459 opportunities that benefit people, society and the environment in East Africa. *Journal of*
460 *Applied Ecology*, 56, 274-281.

461 Ramirez, F., Sbragaglia, V., Soacha, K., Coll, M., & Piera, J. (2022). Challenges for marine
462 ecological assessments: completeness of findable, accessible, interoperable, and reusable
463 biodiversity data in European seas. *Frontiers in Marine Science*, 8, 802235.

464 Roy, H., Groom, Q., Adriaens, T., Agnello, G., Antic, M., Archambeau, A. S., ... & Cardoso, A.
465 C. (2018). Increasing understanding of alien species through citizen science (Alien-
466 CSI). *Research Ideas and Outcomes*, 4, e31412.

- 467 Stork, N. E. (1997). Measuring global biodiversity and its decline. *Biodiversity II:*
468 *understanding and protecting our biological resources*, 41, 41-68.
- 469 Toivonen, T., Heikinheimo, V., Fink, C., Hausmann, A., Hiippala, T., Järv, O., ... & Di Minin, E.
470 (2019). Social media data for conservation science: A methodological overview. *Biological*
471 *Conservation*, 233, 298-315.
- 472 WWF. (2022) Living Planet Report 2022 – Building a naturepositive society. Almond, R.E.A.,
473 Grooten, M., Juffe Bignoli, D. & Petersen, T. (Eds). WWF, Gland, Switzerland.

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List of Figures

475 **Figure 1.** (A) The data extraction pipeline for obtaining species distribution records from
476 Facebook, and (B) the inclusion/exclusion criteria when filtering species photographs for
477 different animal groups of Bangladesh.

478 **Figure 2.** The number and membership details of the available Facebook groups for
479 Bangladeshi animals. Here, 'Herpetofauna' is 'Amphibians' and 'Reptiles', and we named
480 'Multiple taxa' when the Facebook group was focused on > 2 taxa.

481 **Figure 3.** (A) The density (at 1 km²) and (B) yearly growth of species occurrence records,
482 obtained from Facebook for different animal groups of Bangladesh.

483 **Figure 4.** The relative proportion of threatened and non-threatened species in occurrence
484 records (A) and species records (B), obtained from Facebook for different animal groups in
485 Bangladesh, namely amphibians, birds, butterflies, mammals and reptiles. Here the black
486 dots represent the percentages of species that are nationally threatened.

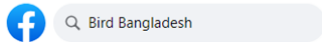
487

488

Step 1. Searching for Facebook groups



1. Search using keyword (e.g., Bird Bangladesh)



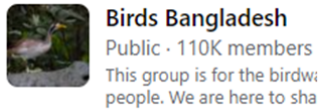
2. Refine search to 'Groups'

Search results

Filters



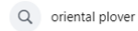
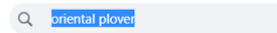
3. Select the relevant



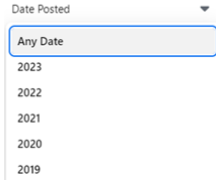
Step 2. Data extraction



1. Search within the group by species name



2. Filter year by selecting 'Date Posted'



3. Check the caption of the photograph and the comments section
4. Check the inclusion/exclusion criteria
5. Extract relevant information (e.g., species)



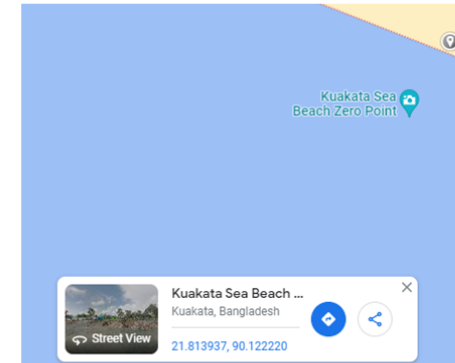
Step 3. Georeferencing the location



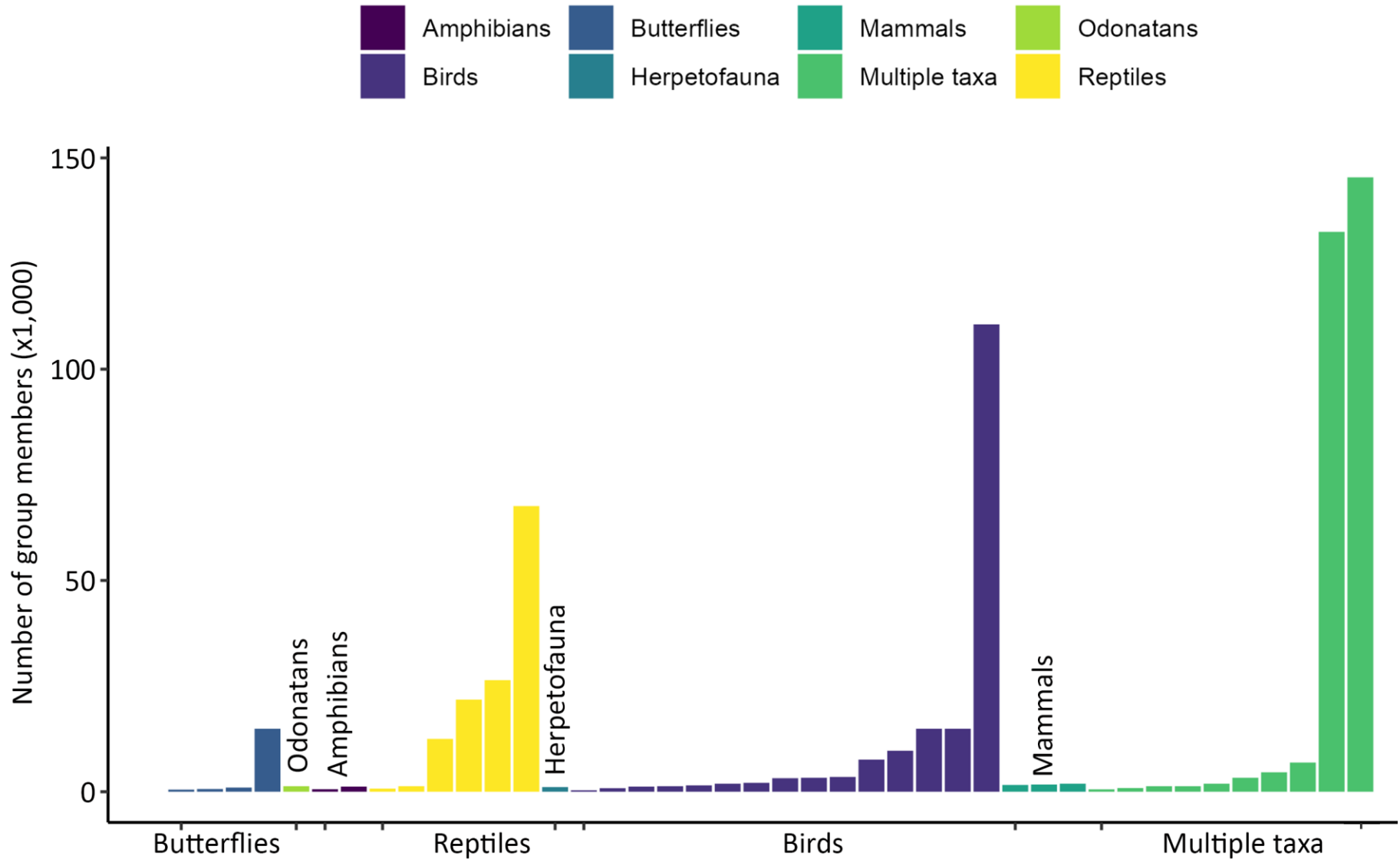
1. Go to Google Map (<https://maps.google.com/>)
2. Search for the location



3. Click the cursor inside the area
4. Extract the latitude and longitude

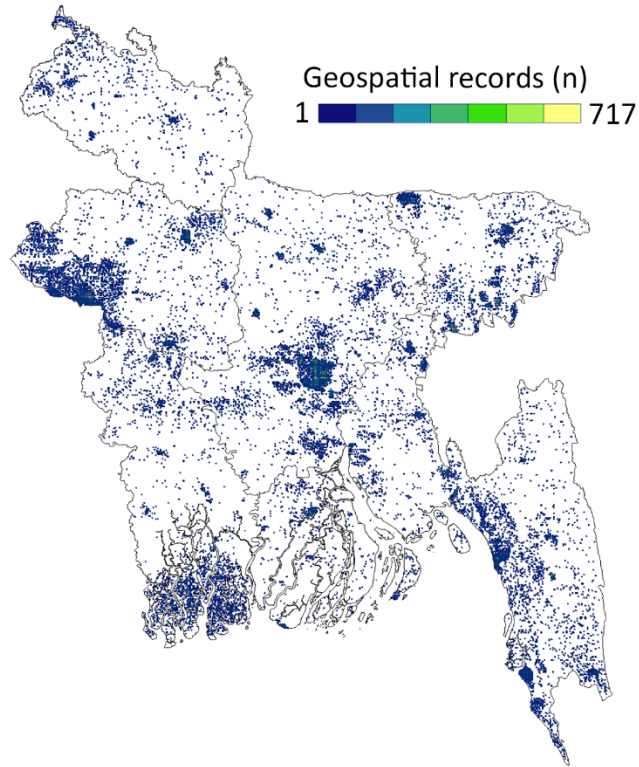


492 **Figure 2.**

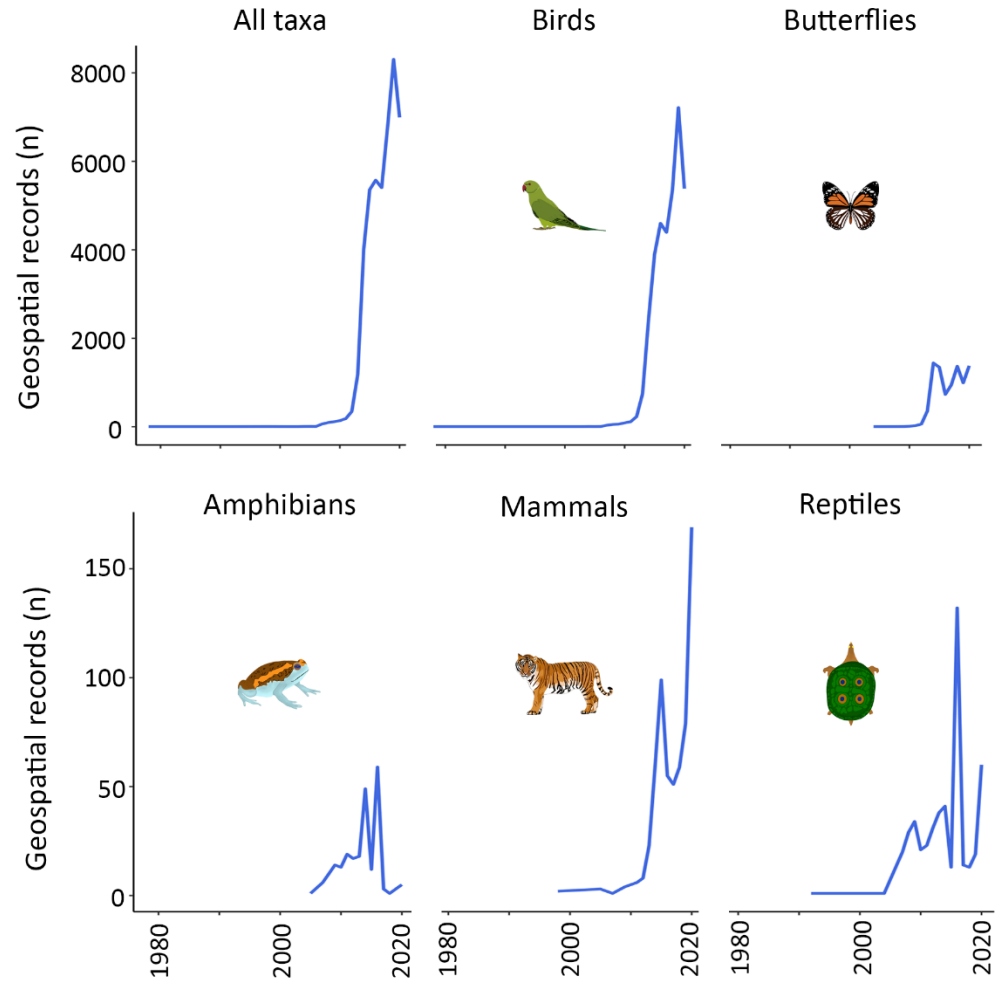


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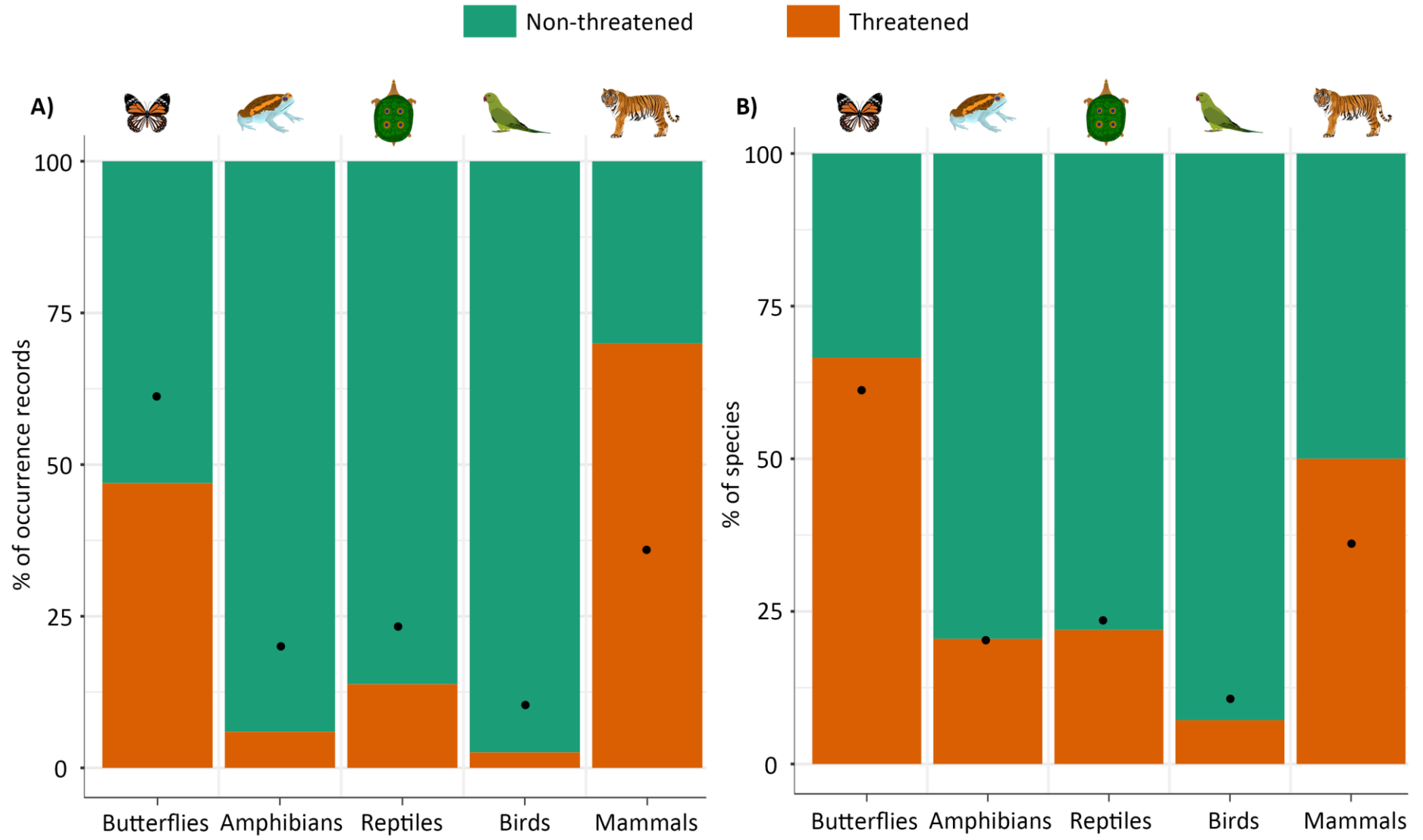
A) Distribution of species occurrence records



B) Annual number of of species occurrence records



496 **Figure 4.**



497