

1 **Climate change impacts on non-human primates: What have we**
2 **modelled and what do we do now?**

3
4 **by**

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41 **ABSTRACT**

42 Climate change will be a key influence on primates in the twenty-first century, potentially
43 exacerbating the effects of habitat loss and anthropogenic activities to drive vulnerable
44 species closer to extinction. There are many ways to assess species' vulnerability to climate
45 change, including modelling approaches of three main types: trait-based models, species
46 distribution models and mechanistic models. In this chapter, we survey the literature on
47 climate change models as applied to primates, including the type(s) of model made and the
48 predictions obtained. Most primate genera (62 of 80) have been subject to ecological
49 modelling, though there are no future projections for lemurs and no palaeoclimate models for
50 lorises, tarsiers or platyrrhines. Maximum entropy methods predominate even though direct
51 comparisons have shown that these tend to predict more severe habitat losses when used
52 uncritically. Most of the taxa modelled to date have been predicted substantial habitat losses
53 by 2100, with significant variation within each taxonomic group.

54

55 **KEYWORDS:** climate change, modelling, anthropogenic impacts, ecological niches,
56 conservation, habitat loss, distributions, species distribution model

57 **1. INTRODUCTION**

58 Climate change is expected to become a major factor driving human and non-human primate
59 survival in the Anthropocene. While climate change impacts may never exceed those of land-

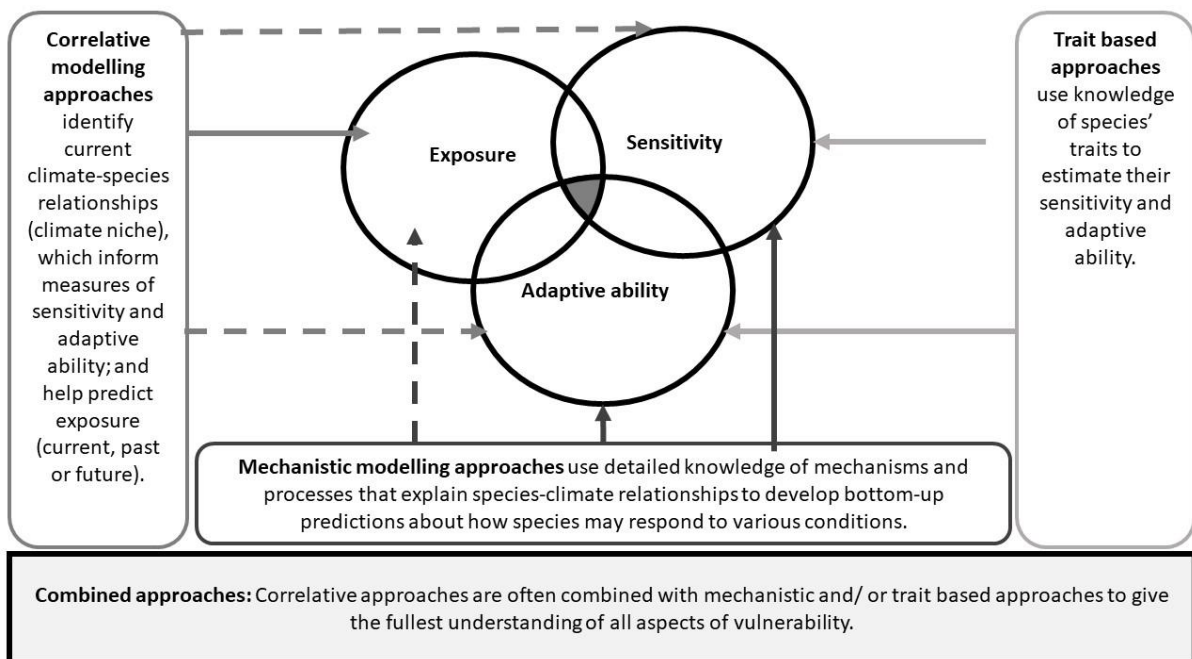
60 use changes and trade, they will very likely exacerbate the effects of these other threats
61 (Gouveia et al., 2016; Struebig et al., 2015 Titeux, 2017; Korstjens and Hillyer 2016).
62 Therefore, a discussion about the effect of anthropogenic factors on non-human primate
63 (henceforth ‘primate’) survival is not complete without some consideration of climate
64 change. Primate vulnerability to climate change is determined by the level and type of
65 changes the species will experience in its geographical range (exposure) coupled with the
66 biological traits that determine how well the species can cope with (sensitivity) or adapt to
67 (adaptability) the predicted changes (Foden et al. 2019; Foden and Young 2016; Korstjens
68 and Hillyer 2016).

69 Many regions where primates occur are expected to undergo pronounced climatic changes
70 (Graham et al. 2016, Carvalho et al. 2019, Zhang et al. 2019a), with increased maximum
71 temperatures, reduced or excessive rainfall, increased seasonality, and an increased frequency
72 and intensity of extreme events (cyclones and droughts) likely to be the biggest threats to
73 primates. Across primate-occupied regions, mean temperature increases are predicted to be
74 10% greater than the global average (i.e. for every 1°C increase globally, primate ranges will
75 heat up by 1.1°C; Graham et al. 2016). More worryingly, maximum temperatures are likely
76 to increase by >2°C across primate ranges, with some areas suffering up to 5-7°C increases,
77 taking these temperatures well above the physiological limit for primates (Carvalho et al.
78 2019). Greatly increased rainfall is predicted for 8% of primate species whilst 4% will see
79 large reductions (>3% increase or decrease per 1°C mean global warming respectively;
80 Graham et al. 2016). Seasonality, an important predictor of species’ geographical ranges
81 (Williams et al. 2021in press), is predicted to become more extreme across primate ranges.
82 Temperature averages, rainfall patterns and seasonality are major determinants of the
83 vegetation cover and productivity of primate habitats. Climate change can exacerbate
84 droughts and cyclones which affect 26% and 18% of primate species respectively (Zhang et
85 al. 2019a) and can lead to devastating instant mortality (Campos et al. 2020). Finally, climate
86 change may have indirect effects on primates by (for instance) changing the patterns of other
87 anthropogenic impacts on them and their habitats. We are still far from producing an
88 exhaustive list of the ways climate change can alter primate lives.

89

90 **1.1 ASSESSING SPECIES VULNERABILITY TO CLIMATE CHANGE:** 91 **MODELLING APPROACHES**

92 Models aimed at assessing species' vulnerability to climate change can be divided into three
 93 categories of increasing complexity, each with a different suitability for assessing sensitivity,
 94 exposure and adaptive ability (Foden et al. 2019; Korstjens and Hillyer 2016; Figure 1). First,
 95 trait-based approaches use expert opinion and published knowledge to determine how a
 96 species is likely to respond to or cope with particular climate changes based on their
 97 biological traits and inferred relationships between traits and vulnerability (e.g. Zhang et al.
 98 2019a). Second, correlative approaches (i.e. species distribution models often called niche or
 99 climate envelope models) investigate how current species distributions are influenced by
 100 climatic conditions and use these relationships to predict how species will be distributed
 101 under future climatic conditions. Finally, mechanistic approaches look at the underlying traits
 102 and physiological processes (mechanisms) that determine environment-species relationships
 103 to predict how species will be able to cope with and respond to changes based on physiology,
 104 behaviour or time budgets (e.g. Dunbar et al. 2009).



105

106 Figure 1: Simplified schematic representation of the relationships between the three main components of a
 107 species' vulnerability and the types of vulnerability assessment approaches used to estimate species'
 108 vulnerability to climate change. Knowledge of each of these elements (sensitivity, exposure, and adaptive
 109 ability) provide the best estimate of actual vulnerability of species (dark grey overlap area in Venn diagram).
 110 Solid arrows show the most basic outcome of different approaches and dashed lines show further more indirect
 111 outcomes or outcomes achieved by combining the approaches. Foden and Young (2016) provide an extensive
 112 overview of vulnerability indices and modelling approaches.

113 Species distribution models (SDMs) identify which areas are most suitable for a particular
114 taxon, as identified by environmental conditions that describe the species' environmental
115 niche (Elith and Leathwick 2009; Norberg 2019). They best explain exposure but also
116 provide important climate niche information to help understand sensitivity and adaptive
117 ability, often used in trait-based and mechanistic approaches (Figure 1). Once a taxon's
118 environmental niche is established based on current distributions, the models can be used to
119 identify other suitable locations elsewhere or in the past or future. SDMs require input data
120 on the environmental conditions at locations where a species is present to compare against
121 conditions where the species is absent. Due to lack of knowledge of absence, many models
122 are actually developed using presence data and 'pseudo-absences' or background locations,
123 i.e. locations that are randomly selected from a predetermined area (e.g. outside the known
124 distribution area) (Santini et al. 2021; Guillera-Arroita et al. 2015). How these pseudo-
125 absences are selected can greatly affect the reliability of the SDM but is not always given
126 sufficient consideration (for a critical review and test see Santini et al. 2021). The variables
127 that are included in establishing this environmental niche depend on the information and
128 computational resources that are available to the researchers, as well as the research question.
129 Recent exponential growth in accessible global datasets on environmental conditions (e.g.
130 landscape cover, climatic conditions, human footprint) and modelling approaches can make it
131 difficult to keep up with best practice (Araújo et al. 2019; Zurell et al. 2020; e.g. Norberg et
132 al. 2019 review 33 statistical SDM approaches).

133 In this chapter, we systematically survey the literature covering modelling of non-human
134 primate responses to climate change. In particular, we consider the potential presence and
135 manifestation of taxonomic bias, explore the state of knowledge for each major primate
136 group, and summarise predictions of habitat changes under future climate conditions.

137

138 **2. SYSTEMATIC METHOD**

139 We searched the literature for modelling studies of the 80 extant non-human primate genera
140 in the IUCN Red List of Threatened Species v. 2020.3 (IUCN 2020) using the broadest
141 academic database, Google Scholar (Gusenbauer 2019). Search terms included the genus
142 name, "species distribution model" and "ecological niche model", linked with Boolean
143 operators. Where we could find no relevant literature for a genus, we refined the taxonomic

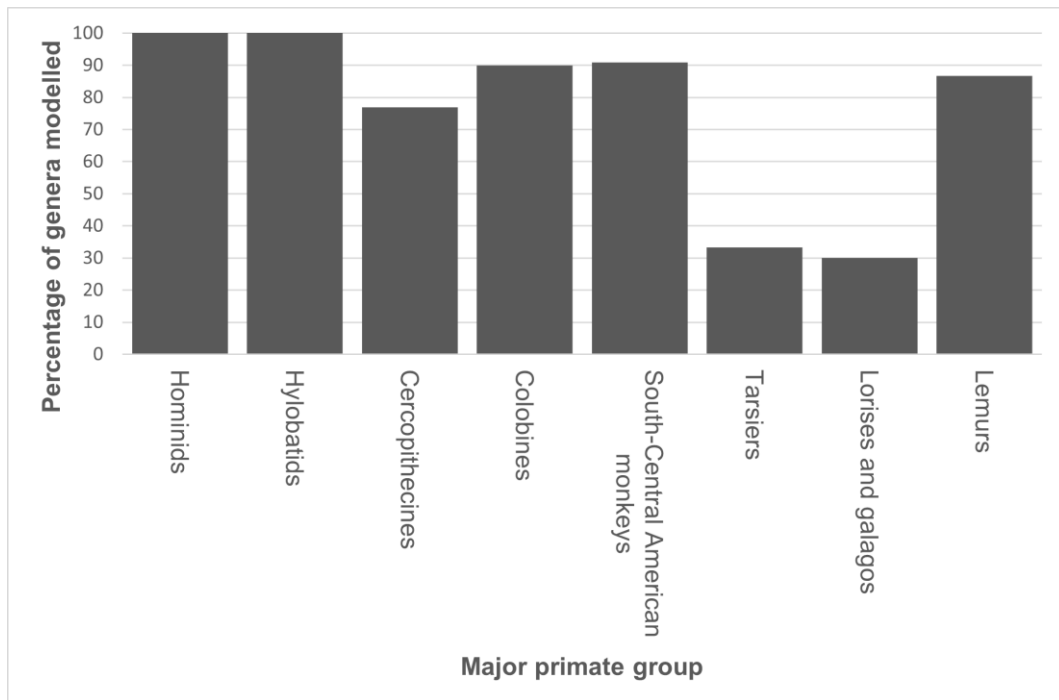
144 search term (e.g. using alternative genus names or adding the rest of the binomial) and finally
145 widened our search using the search term “climate change model”.

146 For each genus we sorted the resulting papers by relevance and selected up to four for
147 detailed analysis (the modal number of relevant hits per genus was 2). Where a genus had
148 more than four relevant papers (which affected the great apes, *Hylobates*, *Macaca*,
149 *Rhinopithecus*, *Cercopithecus*, *Ateles*, *Aotus*, *Alouatta*, *Saguinus* and *Sapajus*), we read
150 abstracts and methods of all we found and selected a representative sample of four that
151 included (1) as many species as possible; (2) a range of approaches, if several had been used;
152 and (3) predictive and/or palaeoclimate studies if they had been done. This meant, for
153 instance, that for the best-studied genera we excluded some very small-scale studies
154 (especially where they modelled a single population or a region rather than a species’ full
155 range). We also sometimes found Masters theses and then papers deriving from the same
156 models (we retained the peer-reviewed papers) or research teams which had produced a series
157 of papers each considering a different species but using the same method and scenario (here
158 we chose a representative example or, if it existed, a recent synthesis covering multiple
159 species). For our final literature sample, we then performed a content analysis by extracting
160 information on climate change scenario(s) and date(s) modelled, modelling approaches, focus
161 and scale (taxon-specific, regional or larger), and the main aims and findings.

162

163 **3. RESULTS**

164 Over half the studies we found dated to 2018-2021. All the major primate groups have been
165 modelled, though coverage ranged from 30-100% of genera (see Figure 2).

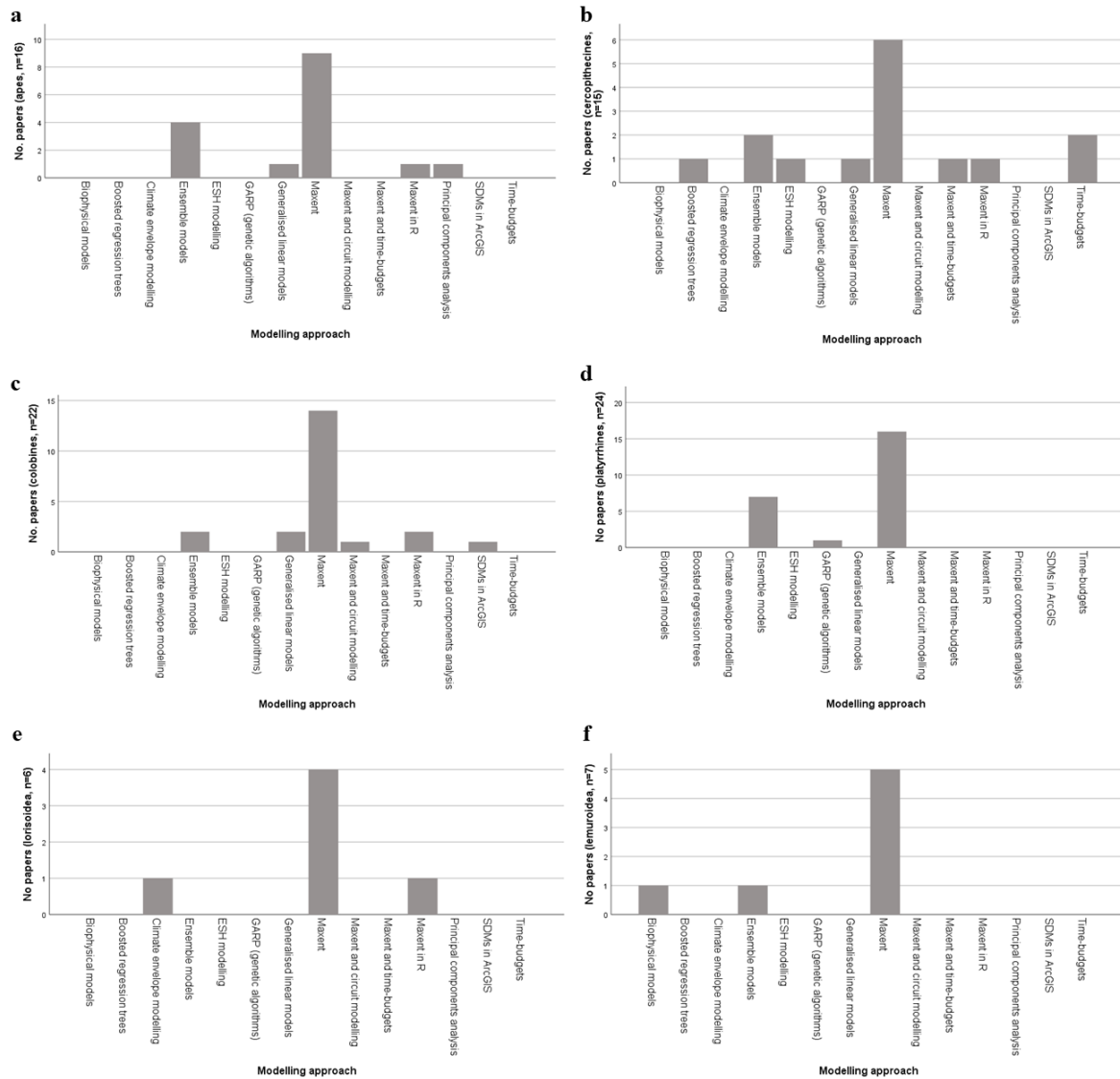


166

167 Figure 2: the percentage of genera that have been modelled, arranged by major taxonomic group.

168 The taxonomic differences in coverage that we observed differed slightly from the bias in
 169 favour of apes and lemurs found in the wider climate change literature (Bernard and Marshall
 170 2020) and are less pronounced than those seen in primatological field studies (Bezanson and
 171 McNamara 2019).

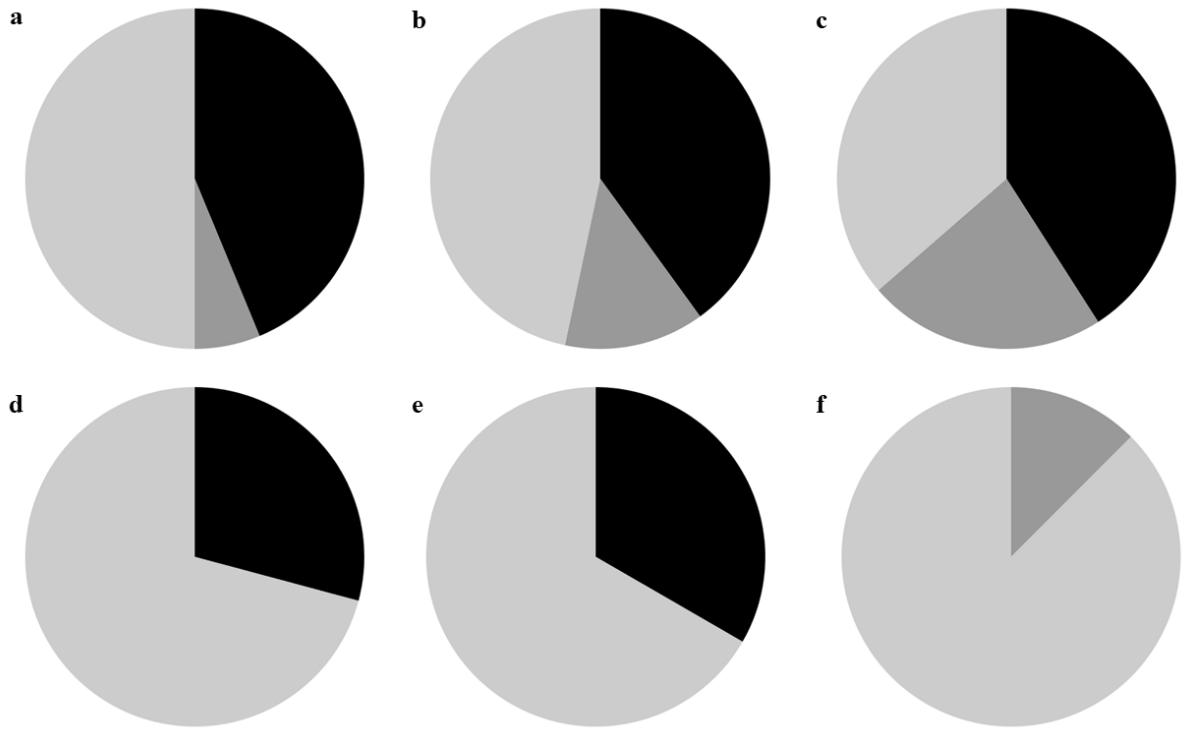
172 Correlative species distribution models dominated our literature (Figure 3). A few studies
 173 used mechanistic time-budget models (Dunbar, 1998 on *Theropithecus*) or trait-based
 174 biophysical models (Stalenberg 2019). The majority of the correlative SDMs used maximum
 175 entropy modelling, either as a stand-alone tool (using Maxent software; Phillips et al., 2006)
 176 or in R (54/89 studies, 65.1%). Ensemble modelling, using multiple approaches (including
 177 maximum entropy, general and generalised linear models) was used in 15 studies (18.1%).
 178 Only 3 (3.6%) used generalised linear models, with all other methods accounting for 1-2.5%
 179 (one or two papers) each. There were some differences among taxonomic groups, with
 180 maximum entropy modelling dominating even more where a group was understudied or
 181 subject to one or two broad works rather than numerous specific ones.



182

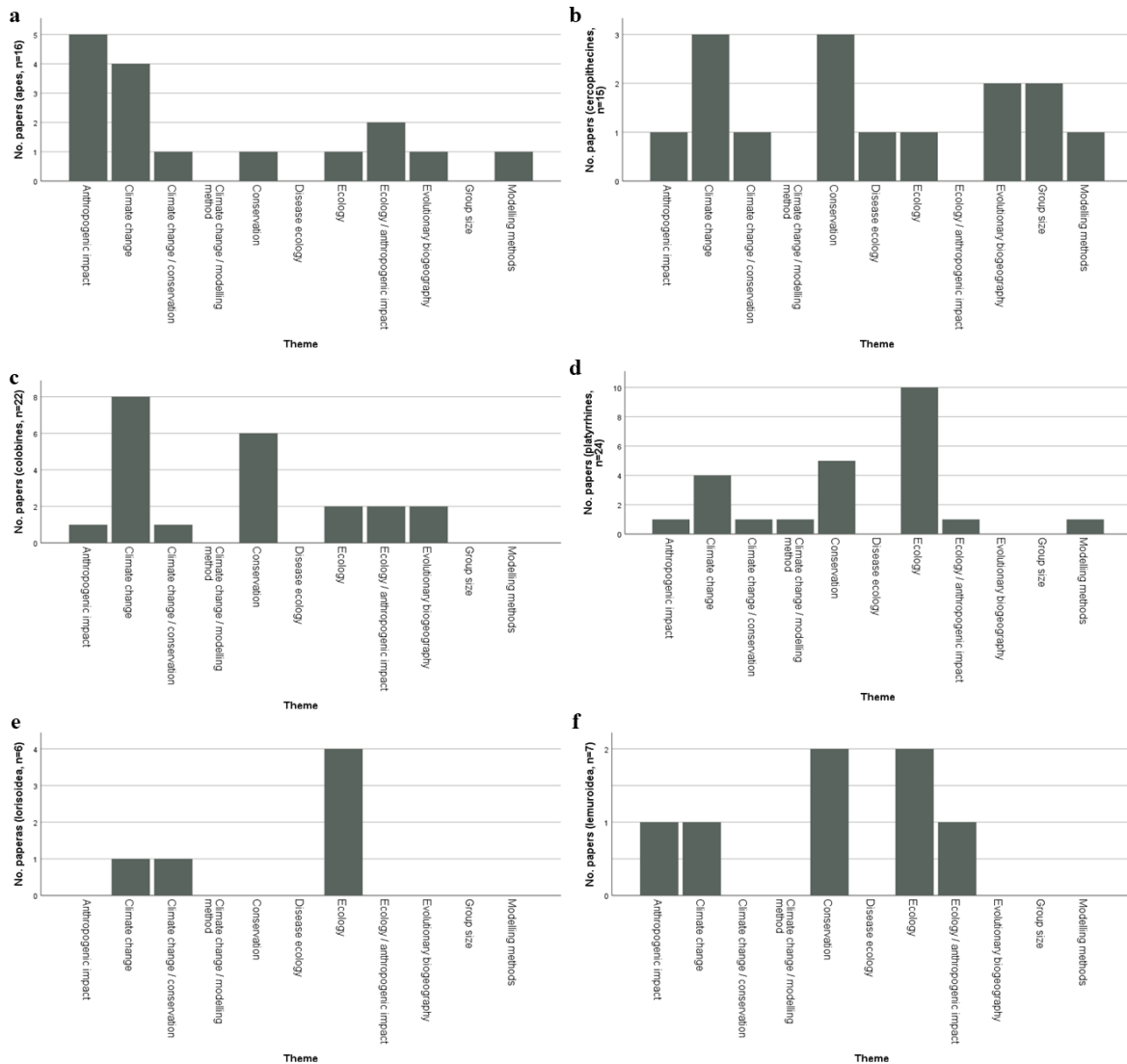
183 Figure 3: modelling approaches used for works on each major primate taxonomic group: a) apes; b)
 184 cercopithecines; c) colobines; d) platyrrhines; e) lorises and galagos; f) lemurs. The tarsiers are excluded,
 185 because we found only one relevant paper, which used maximum entropy modelling in R. Full details of each
 186 paper can be found in supplemental tables 1-7.

187 Most, 54.8% of studies, only modelled the present day, while 33.3% projected into the future
 188 and 10.7% into the past. This pattern varied by group (see Figure 4; there were no future
 189 projections in lemurs; only future but no palaeoclimate models in lorises and galagos and
 190 platyrrhines). Present-only models were generally conservation orientated, looking at
 191 fragmentation, primate distributions and conservation planning. Common themes throughout
 192 the literature include climate change, ecology, anthropogenic impacts, and conservation
 193 (Figure 5).



194

195 Figure 4: periods covered by modelling studies of the major primate groups: a) apes; b) cercopithecines; c)
 196 colobines; d) platyrrhines; e) lorises and galagos; f) lemurs. Black denotes papers projecting the future, mid-grey
 197 include papers examining palaeoclimates, pale grey include papers examining present alone. Tarsiers are
 198 excluded because only one paper (which projected to 2050) was found.



200 Figure 5: common themes in modelling literature for each major primate group: a) apes; b) cercopithecines; c)
 201 colobines; d) platyrrhines; e) lorises and galagos; f) lemurs. Tarsiers are excluded, as only one paper was found
 202 – it explored climate change impacts on biodiversity.

203

204 **3.1. APES**

205 All seven ape genera have been the subject of modelling studies, but only *Hylobates* and the
 206 hominids had >4 studies each. Maximum entropy models were used in 62.5% and ensemble
 207 approaches in 25% of studies. Only Barratt et al. (2020) explored palaeoclimate models, eight
 208 studies modelled the present (Wich et al. 2012; Etiendem et al. 2013; Ario et al. 2018; Singh
 209 et al. 2018; Rahman et al. 2019; Tran and Vu 2020; Bonnin et al. 2020; Yuh et al. 2020) and
 210 seven predicted the future (named below).

211 Outcomes of future predictions varied by taxon and approach. Mwambo (2010) and Carvalho
212 et al. (2021) predicted substantial habitat losses for *Pan* and *Gorilla* by 2090. Thorn et al.
213 (2013), however, found that these losses were typical only when using ‘standard’ correlative
214 models, and a limiting factor model predicted minimal change for mountain gorillas.
215 Predictions of climate change impacts, they concluded, are sensitive to the assumptions of the
216 selected method.

217 For Asian apes, predictions were more negative. Wich et al. (2016) used land-cover change
218 scenarios to predict substantial population declines for *Pongo abelii* by 2030. Other taxa
219 potentially at high risk included various species of *Hylobates* (Trisurat 2018; Condro et al.,
220 2021), *Pongo abelii* and *Symphalangus syndactylus* (Condro et al., 2021). Potentially better
221 off were *Pongo pygmaeus* (Condro et al., 2021) and *Hoolock hoolock* (Deb et al., 2019), both
222 predicted range increases, while *Hylobates muelleri* and *H. abbotti* were predicted only small
223 changes (Condro et al., 2021). The differences in apparent risk may result from different
224 vulnerability profiles or modelling approaches.

225

226 3.2 AFRICAN AND ASIAN MONKEYS

227 We found models on 10/13 cercopithecine genera (all save *Erythrocebus*, *Miopithecus* and
228 *Allenopithecus*, though *Chlorocebus* is included only under the older name *Cercopithecus* in
229 Willems & Hill 2009) and 9/10 colobine genera (all but *Procolobus*). Among the 15 studies
230 found, two predicted past distributions (Khanal et al. 2018a; Chala et al. 2019), six predicted
231 the future (listed below) and seven focused on the present (Willems & Hill 2009; Green
232 2012; Cronin et al. 2015 and 2017, who reported the same models; Moyes et al. 2016;
233 Korstjens et al. 2018; Fuchs et al. 2018; Greenspan et al. 2020).

234 Predicted outcomes of climate change varied. Dunbar (1998)’s time-budget model predicted
235 that *Theropithecus gelada* populations will fragment as climates warm), while Hill and
236 Winder (2019) predicted that 3/6 *Papio* species might suffer substantial habitat losses.
237 Condro et al. (2021) predicted habitat losses by 2050 for eight macaque species, but
238 suggested *Macaca ochreata* and *M. siberu* would be less affected. *Mandrillus leucophaeus*
239 may experience substantial habitat loss from earlier on in the 21st century than *Cercocebus*
240 *atys* (Baker and Willis, 2015). Finally, Korstjens (2019) suggested that habitat suitability for
241 *Cercopithecus* would reduce only slightly by 2070. The final study that looked to the future,
242 Ayebare et al. (2013), did not present any species-specific results.

243 In the colobine literature, there are proportionally more single-species or regional studies than
244 for cercopithecines, particularly in Asia. Of 22 studies, five focused on the past (Moody
245 2007; Winyoningrum 2013; Ehlers-Smith 2014; Ren et al. 2017; Khanal et al. 2018c), nine
246 predicted the future (see below) and eight focused on the present (McDonald et al. 2019;
247 Cronin et al. 2015 and 2017; Cavada et al. 2017; Singh et al. 2018; Tran et al. 2018; Anh et
248 al. 2019; Atmoko et al. 2020; Khanal et al. 2018b).

249 Korstjens (2019) suggested that *Colobus* may be more at risk than *Cercopithecus*. At species
250 level, Baker and Willis (2015) predicted substantial habitat loss for *C. polykomos* (whilst *C.*
251 *vellerosus*, may be unaffected along with *Piliocolobus pennantii*). *P. badius*, however, is
252 predicted range loss by all their models and *P. preussi* may lose habitat by 2040.

253 In Asia, *Pygathrix* (Vu et al., 2020; Tran et al., 2020), *Rhinopithecus* (Luo et al., 2015; Zhang
254 et al., 2019b) and *Semnopithecus* species (Bagaria et al., 2020) are predicted substantial
255 habitat loss. Condro et al. (2021) predicted complete habitat losses for two species of
256 *Trachypithecus*, *Simias concolor* and six species of *Presbytis*. They found that two more
257 *Presbytis* species would experience smaller habitat losses, while three would be stable or
258 increase their habitats. Finally, Zhao et al. (2019) suggested that *Rhinopithecus bieti* might
259 lose 8-22% of its suitable habitat by 2050 unless anthropogenic land-use changes were
260 controlled, in which case, their suitable habitat might increase.

261

262 **3.3 SOUTH AND CENTRAL AMERICAN MONKEYS**

263 We found models for 20 of 22 platyrrhine genera (Platyrrhini), all save *Leontocebus* and
264 *Brachyteles*. Of these, none modelled the past, seven the future (see below) and seventeen the
265 present (Holzmann et al. 2015; Vidal-Garcia and Serio-Silva 2011; Calizto-Peréz et al. 2018;
266 Clément et al. 2014; Hasui et al. 2017; Helenbrook & Valdez 2020; Shanee et al. 2015;
267 Ortega Huerta 2007; Boubli & De Lima 2009; Guy et al. 2016; Moraes et al. 2019; Campos
268 and Jack 2013; Rezende et al. 2020; Ochoa-Wuintero et al. 2017; Garbino et al. 2015;
269 Howard et al. 2015; De Marco et al. 2020).

270 For platyrrhines as for colobines and cercopithecines, projections of climate change impacts
271 often also include the impact of deforestation and other anthropogenic changes, and studies
272 that expressly separate these often suggest forest loss will be more damaging. *Lagothrix*
273 *lagothricha* is predicted a 13% loss of suitable habitat due to climate change, but 18% where
274 forest loss continues apace (Linero et al., 2020). Other species predicted severe habitat losses

275 included *Callithrix flaviceps*, *C. pencillata* and *C. aurita* (though other members of the same
276 genus were less affected; Braz et al., 2019), *Alouatta belzebul*, *Sapajus flavius* and *S.*
277 *libidinosus* (Moraes et al., 2020), *Aotus miconax*, *Lagothrix flavicauda* and, if land-use
278 change continues alongside climate change, *Plecturocebus oenanthe* (Shanee, 2016).

279 Several platyrrhine studies focused on broad biodiversity measures and how biodiversity
280 might shift or decline as climates change. Sales et al. (2017) found that biodiversity of taxa
281 including primates was likely to decline in the Amazon, and Sales et al. (2020) suggested that
282 overall, the contribution of primates to seed-dispersal in this area, based on modelled range
283 area and dispersal potential, is likely to decline substantially. Sales et al. (2019) found that
284 20% of 80 primate species might expand their ranges as a result of climate change, if they can
285 disperse as well as they can today. If fragmentation prevents dispersal of platyrrhines through
286 the Amazon, these 80 species will lose on average 90% of their suitable habitat.

287

288 **3.4 TARSIIERS**

289 Only one of three tarsier genera (genus *Tarsius*) has been modelled, and only once. Condro et
290 al. (2021) used maximum entropy modelling in R to explore climate change impacts on
291 Indonesian biodiversity, and included 8 species. They predicted total habitat loss by 2050 for
292 six species, near complete loss for *Tarsius dentatus* and roughly stable habitat for *T. pumilus*.

293

294 **3.5 LORISES AND GALAGOS**

295 The lorises and galagos (superfamily Lorisioidea) were also rarely studied, with only one of
296 six genera of Galagidae and two of four among Lorisidae the subject of any models. We
297 found no palaeoclimate models, two that predicted the future (see below) and four that
298 focused on the present (Thorn et al. 2009; Nekaris and Stengel 2013; Voskamp et al. 2014;
299 Kumara et al. 2021).

300 Condro et al. (2021) predicted total habitat loss for *N. javanicus*, and ~75% loss for *N.*
301 *bacanus*. The remaining five taxa (*N. kayan*, *N. hilleri*, *N. borneanus*, *N. coucang* and *N.*
302 *menagensis*) were all predicted small increases of up to ~25% of the current extent (Condro
303 et al., 2021). Erasmus et al. (2002) suggested significant shifts in overall biodiversity of
304 South African animal communities, including *Galago moholi*, by 2050.

305

306 3.6 LEMURS

307 We found seven works specifically focusing on modelling the niches and climatic
308 vulnerabilities of Lemuroidea, two of which (Peacock, 2011 and Herrera et al., 2018) cover a
309 wide range of taxa. In fact, beyond these two papers, only *Lepilemur* (Stalenberg, 2019) and
310 *Eulemur*, which was by far the best studied, were modelled. *Allocebus* was not studied at all.
311 Among these, none addressed the future, one (Stalenberg 2019) reconstructed the past and six
312 (Peacock 2011; Herrera et al. 2018; Blair et al. 2013; Kamilar and Tecot 2016; Mercado
313 Malabet et al. 2020; Ormsby 2019) explored the present.

314 The only model of the effects of climate change in lemurs was Stalenberg (2019)'s
315 biophysical model. Biophysical models estimate a species' thermal niche and water
316 requirements and compare physiological needs to environmental conditions to identify
317 suitable habitats for it. Stalenberg's model found that *Lepilemur leucopus* was at higher risk
318 of thermal stress in the 1975-2005 than the 1931-1960 time period.

319

320 4. DISCUSSION AND CONCLUSIONS

321 Climate change impact on primates is still a relatively new area of research, with over half of
322 the studies published 2018 onwards, yet it is no doubt an important one. Our systematic
323 review shows that there are still relatively few studies out there, and fewer than half of the
324 species distribution models that have been built have been used to predict the future. Efforts
325 are not equally distributed across genera, with apes and large-bodied monkeys being the best
326 represented, a bias that we see in the wider primatological literature (Bezanson and
327 McNamara, 2019; Hawes et al., 2013). Bezanson and McNamara (2019) suggest that
328 taxonomic bias may be related to the presence of long-term field sites and publication bias.
329 Although modelling work is often desk-based, it does rely on a reasonable knowledge of the
330 localities where the target species is found.

331 In the wider context of publishing in primatology, the small study counts we found for this
332 specific topic highlight it as an understudied subject. To date, we have very limited
333 understanding of how climate change is potentially going to affect primates, and with 2050
334 (one of the predicted scenarios used in climate modelling) less than 30 years away, we
335 urgently need to make meaningful predictions and adapt subsequent conservation strategies.
336 Those studies we do have suggest that primates will need to shift ranges to accommodate
337 changing climatic conditions (Estrada et al., 2017; Schloss et al., 2012), with consequences

338 for their management and conservation. Primate responses to changing climates will be
339 species-specific (Condro et al., 2021; Schloss et al., 2012 and this study), so we cannot rely
340 on generalisations across subfamilies. The sooner we identify the potential impacts of climate
341 change on primates, the longer we give ourselves to implement the necessary steps to
342 mitigate effects and avoid population declines.

343 Modelling efforts did not seem to correspond with extinction risk, i.e. species classified by
344 the IUCN as “critically endangered” or “endangered” were no more likely to be the focus of
345 these studies. In fact, some endangered species (*Prolemur simus*, *Macaca pagensis*) have yet
346 to be studied this way. Species that are already struggling may face even more challenging
347 futures that we cannot accurately prepare for and mitigate if we do not have predictions of
348 what the future may mean. This is yet again a pattern in the wider literature (Bezanson and
349 McNamara, 2019) and may be a result of data scarcity. Although modelling can draw on
350 online or published pools of data, such as the Global Biodiversity Information Facility
351 (GBIF) or iNaturalist, sometimes datasets for a species are still missing, small, inaccurate or
352 of poor quality (Maldonado et al., 2015), making it difficult to build accurate, high-quality
353 models.

354 Both this review and the literature within it emphasise the need for more foundational
355 knowledge on many primate species. When we are unsure on current ranges, habitat
356 preferences, and a species’ ability to cope with anthropogenic factors, accurate and
357 meaningful projections may be impossible and we run the risk of overgeneralisation. Some
358 species may gain incidental protection from sharing their habitats with others we have
359 modelled, but this is not the ideal given that responses are so species-specific (Schloss et al.,
360 2012; Pacifici et al., 2017). It is also important to note that simply modelling distributions
361 based on perceived correlations between occurrence and climate conditions may not give us
362 the full picture. Herrera et al.’s (2018) study found that food tree distribution was a more
363 accurate predictor for lemur distribution than climate. Indeed, different aspects of the
364 environment will influence different species, with arboreal species perhaps more concerned
365 with forest connectivity than terrestrial ones, and these factors can also affect population
366 density (Pozo-Montuy et al., 2011).

367 Correlative methods, and specifically maximum entropy models, were by far the most
368 common approach we found, but climate change modellers are far from reaching consensus
369 on methodology. With SDMs there is ongoing disagreement on how small a sample size is
370 appropriate and how accurate small-sample studies are (Wisiz et al., 2008; Santini et al.,

2021). Occurrence data used in studies of rare or cryptic species in particular is likely to have been opportunistically collected (and thus not randomly sampled or reflective of entire populations). Conducting studies on the species that have established field sites simply because they are the taxa we have sufficient high-quality data for is likely to unintentionally reinforce taxonomic biases. Additionally, animals do not exist in closed systems. Many external factors that are not commonly included in SDM studies can influence primate distributions. For example, anthropogenic and biotic factors (such as natural disasters) ultimately will play a role in shaping current and future distributions (Graham et al., 2016; Kamilar and Tecot, 2016) and, while some primatological studies included these factors, this was rare. Studies that fail to include anthropogenic layers risk over-estimating primate ranges (Kamilar and Tecot, 2016). SDMs also do not take into account behavioural flexibility, or the issues of sensitivity and adaptability mentioned in our introduction. As a more positive note, it is worth our recognising that some primate species have adapted fairly successfully to human-dominated landscapes, persisting in urban areas (Aguiar et al., 2014; Jaman and Huffman, 2013). Human encroachment and urbanisation may impact such species less than others. Overall, while the picture emerging from existing models of primate responses to climate change may seem grim, we would like to end by proposing that there are significant gains to be made for primatologists and conservationists if we work to develop better methods and simultaneously generate new knowledge about our models' predictive abilities.

390

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684

685 **Supplemental Tables for**

686

687 **Climate change impacts on non-human primates: What have we**
688 **modelled and what do we do now?**

689

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696

697 **Chapter 6**

698 **In**

699

700 **PRIMATES IN ANTHROPOGENIC LANDSCAPES: EXPLORING PRIMATE**
701 **BEHAVIOURAL FLEXIBILITY ACROSS HUMAN CONTEXTS**

702

703 **Editors: Dr. Tracie McKinney, Dr. Sian Waters and Dr. Michelle Rodrigues**

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709 Supplemental Table 1: modelling studies of hominids and hylobatids including taxonomic focus, modelling
 710 approaches, time-periods and broad theme.

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Etiendem et al. (2013)	<i>Gorilla</i>	Maxent			Anthropogenic impact
Thorn et al. (2013)	<i>Gorilla</i>	Ensemble modelling		✓	Modelling methods
Carvalho et al. (2021)	<i>Gorilla, Pan</i>	Ensemble modelling		✓	Climate change
Yuh et al. (2020)	<i>Gorilla, Pan</i>	Maxent			Anthropogenic impact
Deb et al. (2019)	<i>Hoolock</i>	Maxent		✓	Climate change
Trisurat (2018)	<i>Hylobates</i>	Maxent		✓	Climate change
Ario et al. (2018)	<i>Hylobates</i>	Principal components analysis			Conservation
Singh et al. (2018)	<i>Hylobates</i>	Maxent			Ecology / anthropogenic impact
Condro et al. (2021)	<i>Hylobates, Pongo, Symphalangus</i>	Maximum entropy in R		✓	Climate change
Tran & Vu (2020)	<i>Nomascus</i>	Maxent			Ecology
Bonnin et al. (2020)	<i>Pan</i>	Ensemble modelling			Anthropogenic impact
Mwambo (2010)*	<i>Pan</i>	Maxent		✓	Climate change
Barratt et al. (2020)*	<i>Pan</i>	Ensemble modelling	✓		Evolutionary biogeography
Rahman et al. (2019)	<i>Pongo</i>	Maxent			Anthropogenic impact
Wich et al. (2016)	<i>Pongo</i>	Generalised linear modelling		✓	Anthropogenic impact (forest loss scenarios)

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Wich et al. (2012)	<i>Pongo</i>	Maxent			Ecology / anthropogenic impact

711 *papers labelled with an asterisk are yet to be peer-reviewed (theses, pre-prints or reports for conservation
712 organisations).

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715 Supplemental Table 2: modelling studies of cercopithecines including taxonomic focus, modelling approaches,
716 time-periods and broad theme.

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Cronin et al. (2015, 2017)*	<i>Allochrocebus</i> , <i>Cercopithecus</i> , <i>Mandrillus</i>	Maxent			Anthropogenic impact
Korstjens et al. (2018)	<i>Allochrocebus</i> , <i>Cercopithecus</i>	Time-budgets			Group size
Baker and Willis (2015)*	<i>Cercocebus</i> , <i>Mandrillus</i>	Ensemble modelling		✓	Climate change
Korstjens (2019)	<i>Cercopithecus</i>	Generalised linear models		✓	Climate change
Willems and Hill (2009)	<i>Cercopithecus</i> (includes what we would now call <i>Chlorocebus</i>)	Maxent and time-budgets			Modelling methods
Ayebare et al. (2013)	<i>Cercopithecus</i> , <i>Lophocebus</i>	Maxent		✓	Conservation
Khanal et al. (2018a)	<i>Macaca</i>	Maxent	✓		Evolutionary biogeography
Moyes et al. (2016)	<i>Macaca</i>	Boosted regression tree modelling			Disease ecology
Condro et al. (2021)	<i>Macaca</i>	Maximum entropy in R		✓	Climate change / conservation
Greenspan et al. (2020)	<i>Macaca</i>	Maxent			Conservation
Fuchs et al. (2018)	<i>Papio</i>	Maxent			Ecology
Hill and Winder (2019)	<i>Papio</i>	Maxent		✓	Climate change
Chala et al. (2019)	<i>Papio</i>	Ensemble modelling	✓		Evolutionary biogeography

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Green (2012)*	<i>Rungwecebus</i>	Extent of suitable habitat (ESH) modelling			Conservation
Dunbar (1998)	<i>Theropithecus</i>	Time-budgets		✓	Group size

717 *papers marked with an asterisk are yet to be peer-reviewed (theses, pre-prints or reports for conservation
718 organisations.

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721 Supplemental Table 3: modelling studies of colobines including taxonomic focus, modelling approaches, time-
722 periods and broad theme.

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
McDonald et al. (2019)	<i>Colobus</i>	Maxent			Ecology
Korstjens (2019)	<i>Colobus</i>	Generalised linear models		✓	Climate change
Cronin et al. (2015, 2017)*	<i>Colobus</i> , <i>Ptilocolobus</i>	Maxent			Anthropogenic impacts
Cavada et al. (2017)	<i>Colobus</i> , <i>Ptilocolobus</i>	Generalised linear models			Ecology
Baker and Willis (2015)*	<i>Ptilocolobus</i>	Ensemble modelling		✓	Climate change
Condro et al. (2021)	<i>Presbytis</i> , <i>Nasalis</i> , <i>Simias</i> , <i>Trachypithecus</i>	Maximum entropy in R		✓	Climate change / conservation
Ehlers-Smith (2014)	<i>Presbytis</i>	Species distribution modelling in ArcGIS	✓		Ecology / anthropogenic impact
Singh et al. (2018)	<i>Presbytis</i>	Maxent			Ecology / anthropogenic impact
Vu et al. (2020)	<i>Pygathrix</i>	Maxent		✓	Climate change
Tran et al. (2020)	<i>Pygathrix</i>	Maxent		✓	Climate change
Tran et al. (2018)	<i>Pygathrix</i>	Maxent			Conservation
Anh et al. (2019)	<i>Pygathrix</i>	Maxent			Conservation
Atmoko et al. (2020)	<i>Nasalis</i>	Maxent			Conservation
Zhao et al. (2019)	<i>Rhinopithecus</i>	Maxent and genetic/ circuit modelling		✓	Climate change

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Luo et al. (2015)	<i>Rhinopithecus</i>	Maxent		✓	Climate change
Zhang et al. (2019b)	<i>Rhinopithecus</i>	Maxent		✓	Climate change
Ren et al. (2017)	<i>Rhinopithecus</i>	Maximum entropy in R	✓		Conservation
Khanal et al. (2018b)	<i>Semnopithecus</i>	Maxent			Conservation
Khanal et al. (2018c)	<i>Semnopithecus</i>	Maxent	✓		Evolutionary biogeography
Bagaria et al. (2020)	<i>Semnopithecus</i>	Ensemble modelling		✓	Climate change
Moody (2007)*	<i>Trachypithecus</i>	Maxent	✓		Evolutionary biogeography
Windyoningrum (2013)*	<i>Trachypithecus</i>	Maxent	✓		Conservation

723 *papers marked with an asterisk are yet to be peer-reviewed (theses, pre-prints or reports for conservation
724 organisations).

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727 Supplemental Table 4: modelling studies of platyrrhines including taxonomic focus, modelling approaches,
 728 time-periods and broad theme.

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Holzmann et al. (2015)	<i>Alouatta</i>	Maxent			Ecology
Sales et al. (2017)	<i>Alouatta, Aotus, Ateles, Callicebus, Cebuella, Chiropotes, Saguinus, Saimiri</i>	Ensemble modelling		✓	Climate change / modelling methods
Vidal-Garcia and Serio-Silva (2011)	<i>Alouatta, Ateles</i>	Maxent			Ecology
Calixto-Pérez et al. (2018)	<i>Alouatta, Ateles</i>	Maxent			Modelling method
Clement et al. (2014)	<i>Alouatta, Ateles, Cebus, Pithecia, Saguinus, Saimiri</i>	Ensemble modelling			Ecology
Sales et al. (2019)	<i>Alouatta, Aotus, Ateles, Cacajao, Callibella, Callimico, Cebuella, Cebus, Cheracebus, Chiropotes, Lagothrix, Mico, Pithecia, Plecturocebus, Saguinus, Saimiri, Sapajus</i>	Maxent		✓	Climate change

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Hasui et al. (2017)	<i>Alouatta</i> , <i>Callicebus</i> , <i>Callithrix</i> , <i>Sapajus</i>	Maxent			Ecology
Moraes et al. (2020)	<i>Alouatta</i> , <i>Sapajus</i>	Maxent		✓	Conservation
Shanee (2016)	<i>Aotus</i> , <i>Lagothrix</i> , <i>Plecturocebus</i> ,	Maxent		✓	Climate change
Helenbrook and Valdez (2020)*	<i>Aotus</i>	Maxent			Conservation
Shanee et al. (2015)	<i>Aotus</i>	Maxent			Ecology
Ortega Huerta (2007)	<i>Ateles</i>	GARP (genetic algorithms)			Conservation
Sales et al. (2020)	<i>Ateles</i> , <i>Cebus</i> , <i>Cheracebus</i> , <i>Lagothrix</i> , <i>Pithecia</i> , <i>Plecturocebus</i> , <i>Saguinus</i> , <i>Saimiri</i> , <i>Sapajus</i>	Maxent		✓	Climate change
Boubli and De Lima (2009)	<i>Cacajao</i> , <i>Chiropotes</i>	Maxent			Ecology
Guy et al. (2016)	<i>Callithrix</i> , <i>Leontopithecus</i>	Maxent			Ecology
Moraes et al. (2019)	<i>Callithrix</i>	Ensemble modelling			Anthropogenic impact
Braz et al. (2019)	<i>Callithrix</i>	Ensemble modelling		✓	Climate change
Campos and Jack (2013)	<i>Cebus</i>	Maxent			Conservation

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Linero et al. (2020)	<i>Lagothrix</i>	Ensemble modelling		✓	Climate change / conservatopn
Rezende et al. (2020)	<i>Leontopithecus</i>	Ensemble modelling			Conservation
Ochoa-Quintero et al. (2017)	<i>Mico</i>	Maxent			Ecology
Garbino et al. (2015)	<i>Saguinus</i>	Maxent			Ecology
Howard et al. (2015)	<i>Sapajus</i>	Maxent			Ecology
De Marco et al. (2020)	<i>Sapajus</i>	Ensemble modelling			Ecology / anthropogenic impact

729 *papers marked with an asterisk are yet to be peer-reviewed (theses, pre-prints or reports for conservation
730 organisations).

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732 Supplemental Table 5: modelling studies of tarsiers including taxonomic focus, modelling approaches, time-
733 periods and broad theme.

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Condro et al. (2021)	<i>Nycticebus</i>	Maximum entropy in R		✓	Climate change / conservation

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736 Supplemental Table 6: modelling studies of lorises and galagos including taxonomic focus, modelling
 737 approaches, time-periods and broad theme.

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Erasmus et al. (2002)	<i>Galago</i>	Climate envelope modelling		✓	Climate change
Nekaris and Stengel (2013)	<i>Loris</i>	Maxent			Ecology
Voskamp et al. (2014)	<i>Nycticebus</i>	Maxent			Ecology
Kumara et al. (2021)	<i>Nycticebus</i>	Maxent			Ecology
Condro et al. (2021)	<i>Nycticebus</i>	Maximum entropy in R		✓	Climate change / conservation
Thorn et al. (2009)	<i>Nycticebus</i>	Maxent			Ecology

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740 Supplemental Table 7: modelling studies of lemurs including taxonomic focus, modelling approaches, time-
 741 periods and broad theme.

Paper	Genera	Modelling approach	Periods		Theme
			Past	Future	
Peacock (2011)	<i>Avahi</i> , <i>Cheirogaleus</i> , <i>Daubentonia</i> , <i>Eulemur</i> , <i>Hapalemur</i> , <i>Indri</i> , <i>Lemur</i> , <i>Lepilemur</i> , <i>Microcebus</i> , <i>Mirza</i> , <i>Phaner</i> , <i>Propithecus</i> , <i>Varecia</i>	Maxent			Conservation
Herrera et al. (2018)	<i>Avahi</i> , <i>Cheirogaleus</i> , <i>Eulemur</i> , <i>Hapalemur</i> , <i>Indri</i> , <i>Lemur</i> , <i>Lepilemur</i> , <i>Microcebus</i> , <i>Propithecus</i> , <i>Varecia</i>	Maxent, with general and generalised linear models to predict densities			Ecology
Blair et al. (2013)	<i>Eulemur</i>	Maxent			Ecology
Kamilar and Tecot (2016)	<i>Eulemur</i>	Maxent			Ecology / anthropogenic impacts
Mercado Malabet et al. (2020)	<i>Eulemur</i>	Maxent			Anthropogenic impact
Ormsby (2019)	<i>Eulemur</i>	Ensemble modelling			Conservation
Stalenberg (2019)*	<i>Lepilemur</i>	Biophysical (mechanistic) models	✓		Climate change

742 *papers marked with an asterisk are yet to be peer-reviewed (theses, pre-prints or reports for conservation
743 organisations).

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745 Additional tables in supplemental material:

746 <https://doi.org/10.18746/bmth.data.00000220>