# Trends and biases in African large carnivore population assessments: Identifying priorities and opportunities from a systematic review of two decades of research (#74907)

First submission

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# Trends and biases in African large carnivore population assessments: Identifying priorities and opportunities from a systematic review of two decades of research

Paolo Strampelli Corresp., Equal first author, 1, 2, Liz A.D. Campbell Equal first author, 1, Philipp Henschel 3, Samantha K. Nicholson 4, 5, David W. Macdonald 1, Amy J. Dickman 1, 2

Corresponding Author: Paolo Strampelli Email address: paolo.strampelli@zoo.ox.ac.uk

African large carnivores have undergone significant range and population declines over recent decades. Although conservation planning and the management of threatened species requires accurate assessments of population status and monitoring of trends, there is evidence that biodiversity monitoring may not be evenly distributed or occurring where most needed. Here, we provide the first systematic review of African large carnivore population assessments published over the last two decades (2000-2020), to investigate trends in research effort and identify knowledge gaps. We used generalised linear models (GLMs) and generalised linear mixed models (GLMMs) to identify taxonomic and geographical biases, and investigated biases associated with land use type and author nationality. Research effort was significantly biased towards lion (Panthera leo) and against striped hyaena (*Hyaena hyaena*), despite the latter being the species with the widest continental range. African wild dog (Lycaon pictus) also exhibited a negative bias in research attention, although this was partly explained by its relatively restricted distribution. The number of country assessments for a species was significantly positively associated with its geographic range in that country. Population assessments were biased towards southern and eastern Africa, particularly South Africa and Kenya. Northern, western, and central Africa were generally under-represented. Most studies were carried out in photographic tourism protected areas under government management, while nonprotected and trophy hunting areas received less attention. Outside South Africa, almost half of studies (41%) did not include authors from the study country, suggesting that significant opportunities exist for capacity building in range states. Overall, large parts of Africa remain under-represented in the literature, and opportunities exist for further

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 $<sup>^{</sup>m 1}$  Wildlife Conservation Research Unit (WildCRU), Department of Zoology, University of Oxford, Oxford, United Kingdom

<sup>&</sup>lt;sup>2</sup> Lion Landscapes, Iringa, Tanzania

Panthera, New York, United States of America

<sup>4</sup> Endangered Wildlife Trust, Johannesburg, South Africa

<sup>&</sup>lt;sup>5</sup> The University of KwaZulu-Natal, Durban, South Africa



research on most species and in most countries. We develop recommendations for actions aimed at overcoming the identified biases and provide researchers, practitioners, and policymakers with priorities to help inform future research and monitoring agendas.



- 1 Trends and biases in African large carnivore population assessments:
- 2 Identifying priorities and opportunities from a systematic review of
- 3 two decades of research
- 4
- $5 \quad \text{Paolo Strampelli}^{1,2,\P,*}, \text{Liz A.D. Campbell}^{1,\P}, \text{Philipp Henschel}^3, \text{Samantha Nicholson}^{4,5}, \text{David Paolo Strampelli}^{1,2,\P,*}, \text{Liz A.D. Campbell}^{1,\P,*}, \text{Philipp Henschel}^3, \text{Samantha Nicholson}^{4,5}, \text{David Paolo Strampelli}^{1,2,\P,*}, \text{Liz A.D. Campbell}^{1,2,\P,*}, \text{Philipp Henschel}^{3,-}, \text{Campbell}^{3,-}, \text{C$
- 6 W. Macdonald<sup>1</sup>, Amy J. Dickman<sup>1,2</sup>
- 7 Paolo Strampelli and Liz A.D. Campbell are joint first authors

- 9 <sup>1</sup> Wildlife Conservation Research Unit, Department of Zoology, Recanati Kaplan Centre,
- 10 University of Oxford, Oxford, UK
- 11 <sup>2</sup> Lion Landscapes, Iringa, Tanzania
- 12 <sup>3</sup> Panthera, New York, NY, USA
- 13 <sup>4</sup> Endangered Wildlife Trust, Johannesburg, South Africa
- <sup>5</sup> The University of KwaZulu-Natal, South Africa

15

- 16 Corresponding Author:
- 17 Paolo Strampelli, Wildlife Conservation Research Unit (WildCRU), Department of Zoology,
- 18 University of Oxford, Tubney House, Abingdon Road, Tubney, Abingdon OX13 5QL, UK
- 19 Email Address: paolo.strampelli@zoo.ox.ac.uk; paolo.strampelli@gmail.com



# **Abstract**

21	African large carnivores have undergone significant range and population declines over recent
22	decades. Although conservation planning and the management of threatened species requires
23	accurate assessments of population status and monitoring of trends, there is evidence that
24	biodiversity monitoring may not be evenly distributed or occurring where most needed. Here, we
25	provide the first systematic review of African large carnivore population assessments published
26	over the last two decades (2000-2020), to investigate trends in research effort and identify
27	knowledge gaps. We used generalised linear models (GLMs) and generalised linear mixed
28	models (GLMMs) to identify taxonomic and geographical biases, and investigated biases
29	associated with land use type and author nationality. Research effort was significantly biased
30	towards lion (Panthera leo) and against striped hyaena (Hyaena hyaena), despite the latter being
31	the species with the widest continental range. African wild dog (Lycaon pictus) also exhibited a
32	negative bias in research attention, although this was partly explained by its relatively restricted
33	distribution. The number of country assessments for a species was significantly positively
34	associated with its geographic range in that country. Population assessments were biased towards
35	southern and eastern Africa, particularly South Africa and Kenya. Northern, western, and central
36	Africa were generally under-represented. Most studies were carried out in photographic tourism
37	protected areas under government management, while non-protected and trophy hunting areas
38	received less attention. Outside South Africa, almost half of studies (41%) did not include
39	authors from the study country, suggesting that significant opportunities exist for capacity
40	building in range states. Overall, large parts of Africa remain under-represented in the literature,
41	and opportunities exist for further research on most species and in most countries. We develop
42	recommendations for actions aimed at overcoming the identified biases and provide researchers,

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43	practitioners.	and police	vmakers v	with	priorities	to hel	p inform	future res	earch and	monitorir

44 agendas.



# 45 Introduction

46	Africa is host to a unique diversity of large carnivore species, including the lion ( <i>Panthera leo</i> ),
47	leopard (Panthera pardus), cheetah (Acinonyx jubatus), African wild dog (Lycaon pictus),
48	spotted hyaena (Crocuta crocuta), striped hyaena (Hyaena hyaena), and brown hyaena
49	(Parahyaena brunnea). Nevertheless, large carnivore populations across Africa have undergone
50	rapid declines in recent decades, primarily as a result of habitat loss and fragmentation, loss of
51	prey, and conflict with humans (Brodie et al., 2021; Wolf and Ripple, 2017). There is therefore
52	an urgent need to ensure that remaining populations are effectively studied and monitored; in
53	addition to being fundamental for understanding population dynamics (Elliot and Gopalaswamy,
54	2016), knowledge of the status and trends of populations is essential for their conservation
55	management, as it allows practitioners to identify threats, evaluate the effectiveness of
56	interventions, implement adaptive monitoring programmes, and inform policy decisions
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57 58 59 60 61 62	(Suryawanshi et al., 2019; Witmer, 2005).  Indeed, such data have been used in recent years to inform a wide range of management and policy decisions affecting large African carnivores, including national and regional action plans and conservation strategies (IUCN, 2007; TAWIRI, 2016), range-wide meta-analyses (Weise et al., 2017), international policies (USFWS, 2015), extinction risk assessments (IUCN, 2020), and trophy hunting quota setting (Mweetwa et al., 2018). They have also been employed to identify
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estimating densities also allows comparison over time and between sites, enabling researchers 67 and managers to understand how population status varies with biotic factors (Searle et al., 2021), 68 anthropogenic disturbances (Balme et al., 2010; Henschel et al., 2011), or land management 69 strategies (Swanepoel et al., 2015). Without such data, knowledge gaps are often filled by expert 70 opinion (Weise et al., 2017), which can delay or prevent conservation actions (Artelle et al., 71 72 2013; Popescu et al., 2016), or lead to inappropriate or harmful management decisions (Darimont et al., 2018; Moganaki et al., 2018). Successful large carnivore conservation therefore requires 73 reliable population assessments (Braczkowski et al., 2020), ideally from a wide range of 74 geographical and management contexts. 75 76 However, as a result of their naturally low densities, nocturnal and secretive nature, and wide ranging habits, estimating population parameters for large carnivores can be logistically 77 challenging and financially costly (Karanth and Nichols, 2017). Indeed, African carnivore 78 79 population measures have been argued to be severely lacking (Riggio et al., 2013), and there is 80 evidence that biodiversity monitoring and research as a whole is under-represented in Africa (Di Marco et al., 2017; Martin et al., 2012; Stocks et al., 2008; Velasco et al., 2015). Conservation 81 research also suffers from both taxonomical and geographical sampling biases, with research not 82 83 always targeting the areas or species with the largest knowledge gaps (Di Marco et al., 2017). Indeed, research can be geographically biased not only at the country level, but also towards 84 specific regions, ecosystems, and land use categories (Velasco et al., 2015). Resulting knowledge 85 gaps can compromise the implementation of science-based conservation interventions (Trimble 86 and van Aarde, 2012), and biases might translate into policies, impacting the achievement of 87 biodiversity conservation targets (Velasco et al., 2015). It is therefore important to understand 88 and monitor such biases in research, in order to track re-align research priorities where needed 89



(Di Marco et al., 2017; Donaldson et al., 2016). Given ongoing funding shortfalls in conservation 90 (Lindsey et al., 2018), it is especially important that limited resources are directed toward where 91 they are most needed (Trimble and van Aarde, 2012). 92 93 As a result of this need, we carry out the first systematic review of all published, peer-review studies estimating population density or abundance of large African carnivore populations over 94 95 two decades (2000 - 2020). We determine research patterns and the geographical and taxonomic representativeness of these studies, and identify biases and data gaps in population assessments 96 97 and monitoring. We also employ findings to discuss representativeness of different land use types in research efforts, and the extent of involvement of authors from host countries. Finally, 98 we use this information to identify research priorities and opportunities, expanding on how future 99 large carnivore population assessments can be best employed to guide conservation management 100 of these species. 101

# Materials & Methods

# Literature review

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We conducted a systematic review of published, peer-reviewed literature using Google Scholar (last search: January 2021), compiling all studies where population density of one or more African large carnivore species (lion, leopard, cheetah, African wild dog, spotted hyaena, striped hyaena, brown hyaena) was either explicitly estimated, or derived from an empirically measured parameter (e.g. home ranges). The review protocol was applied following the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) guidelines (Page et al., 2021; see Fig 1 for PRISMA flow diagram and Appendix S1 for PRISMA checklist). Literature was searched by P.S. and S.N. by entering, in quotation marks, full scientific and vernacular



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species names, as well as "density", "abundance", "population", "assessment", or "survey". As our intention was to review trends and biases in peer-reviewed literature over two decades, only studies published between 1st January 2000 and 31st December 2020 were considered. Due to the number of articles returned using Google Scholar searches, only the first 100 results for each search were considered. P.S. and S.N. screened the resulting records, and all papers that did not include density or abundance estimation as an output were excluded (Fig. 1). There were no disagreements on classifications. Only estimates published in peer-reviewed literature were considered, to ensure data and information quality (Suryawanshi et al., 2019). In addition, due to many unpublished reports not being publicly available, studies published in the scientific literature are those most readily available to most policy makers, conservation managers, and researchers, and therefore most likely to influence international conservation policy and the disbursement of conservation funding (Giam and Wilcove, 2012). Although it was not necessary for population density to be the primary parameter of interest in the study, studies were only included if the methodology employed to obtain density estimates was described, thus providing the opportunity for this aspect of the study to be subjected to peer review. Studies estimating population abundance were included only if the size of the sampled area was explicitly defined and measured. Estimates reported in a peer-reviewed study from nonpeer reviewed sources (e.g. Weise et al., 2017), where the methodology employed was not described (e.g. Balme et al., 2017), or where estimates for a wider area were obtained indirectly by extrapolating from a smaller sampled area (e.g. Trinkel, 2009) were not included. We also did not include density estimates from intensively managed populations (e.g. Buk et al., 2018), as the exact number of individuals being known prior to the study precluded the type of 'exploratory'





surveys of interest in this review. This was only relevant to some populations of lion, cheetah, 135 and African wild dog in South Africa. 136 137 For each study that fitted the above criteria, we extracted information on: year of publication, 138 year(s) of data collection, authors' nationality (national/citizen of study country or foreigner, based on a web search), study area, ecosystem, country, region, land use type (as described in the 139 140 publication; if multiple uses occurred – e.g. photographic tourism and trophy hunting – both were listed), density estimate, and estimation method. Risk of bias was minimised by searching 141 142 for multiple possible terms in Google Scholar and by the clear-cut definitions of inclusion employed 143 Finally, we are aware of recent debates regarding the reliability of some large carnivore 144 145 population assessment methods (e.g. Braczkowski et al., 2020; Dröge et al., 2020), and of common issues in large carnivore density estimation studies, such as that of under-sampling the 146 study area (Suryawanshi et al., 2019). However, given that our goal was to assess research effort, 147 we did not make distinctions based on methods employed, or on other features of the studies 148 themselves. 149 **Analyses** 150 Findings of the literature review were employed to identify taxonomic trends in research effort, 151 including the total number of studies and the study density (studies per km<sup>2</sup> of geographical range) 152 per species. Geographic range maps and data for each species were obtained from the IUCN Red 153 List (https://www.iucnredlist.org/). We also determined spatial patterns in research, through the 154 155 total number of studies and the study density per country and per region, both by species and

overall. Finally, we investigated patterns in land tenure of the areas where research was carried

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out, and the iation and nationality (local national versus foreigner) of the authors of the 157 research. 158 159 We employed Generalised Linear Mixed Models (GLMMs) and Generalised Linear Models 160 (GLMs) to investigate taxonomic and spatial biases in large carnivore population assessments. Models that combined species were fitted by Poisson GLMMs, with the number of studies coded 161 162 as the explanatory variable. Country and Species were included as random effects (random intercepts) to control for multiple observations (i.e. data points for multiple species in each country, 163 and for multiple range countries for each species), and to investigate research biases (see below). 164 All models were validated by posterior predictive checks, dispersion parameters, and model 165 residuals (Appendix S5). 166 167 To first determine overall biases in research effort by species and countries, a model was built with only Species and Country as random effects (Model 1). In this model, the random intercepts 168 represent relative differences in the number of surveys for that species or country; positive values 169 indicate a greater than average number of studies for that species or country, while negative values 170 indicate a lower than average number of studies, with the magnitude indicating the degree of 171 172 erent. Species with larger geographical ranges could be expected to receive greater research attention by 173 chance. Therefore, to control for this, we then fit a model controlling for differences in species' 174 175 geographical range (Range) between countries, with Range modelled as a fixed effect and Country and Species and random effects (Model 2). This allowed us to (a) test whether greater research 176 effort was directed towards species and countries with larger ranges; and (b) assess biases in 177 178 research, by country and by species, after controlling for differences in geographical range. For



this model, the random intercepts indicate whether the number of studies for a species or in a 179 country was more or less than expected according to the range of that species or in that country. 180 181 All analyses followed a Bayesian approach to parameter estimation using JAGS (Plummer, 2003) 182 in R (version 3.6.1; R Core Team, 2019) with the package *runjags* (Denwood, 2020). Explanatory variables were standardised to improve MCMC chain convergence and aid interpretation of results. 183 184 Prior to analyses, variable collinearity was assessed using variance inflation factors and pairwise correlations, which found no issues. All models used three MCMC chains, diffuse priors, and ran 185 186 for enough iterations to produce an effective sample size >10,000 (Kruschke and Liddell, 2018). MCMC chain convergence was confirmed with trace-plots and the Gelman-Rubrin statistic. 187 188 Variables (both fixed and random effects) were considered to have a significant effect if they exhibited a credible non-zero effect (i.e. zero not contained within the 95% highest density interval 189 (HDI) of the posterior distribution). 190 In addition to the all-species GLMMs, GLM were also built for each species to investigate species-191 specific research biases. GLMs were used over GLMMs as – unlike the all-species models – 192 datasets for individual species did not have multiple observations per country, thus allowing their 193 194 applicability. Country biases were assessed using the model residuals: a positive residual signified a country had more studies than expected, while a negative residual signified fewer. Residuals 195 considered credibly different from zero were those for which zero was not contained in the 95% 196 197 highest density interval (HDI). GLMs were first fitted with a Poisson distribution and validated by posterior predictive checks, 198 dispersion parameters, and model residuals. If a Poisson model was not suitable for the data, zero-199 200 inflated Poisson, negative binomial, or zero-inflated negative binomial models were used, validated by the same methods employed for the GLMMs (Appendix S5). Through this method, 201



leopard, spotted hyaena, and cheetah data were fitted with Poisson GLMs with one outlier removed to reduce overdispersion; lion data were fitted with a Negative Binomial GLM. African wild dog, striped hyaena, and brown hyaena could not be modelled individually due to too few data points (range countries) and/or insufficient variance (i.e. most range countries had zero studies).

# Results

### Literature review and research trends

Our search of the published literature revealed 115 peer-reviewed articles (studies) which estimated population density of one or more can large African species. These provided a total of 312 estimates of large carnivore population density, across all species (Fig. 2, 3).

Studies were predominantly carried out on protected land managed by the government (63%), followed by private reserves (19%), land under community-based management (16%), and unprotected areas (10%). Land tenure was unclear for 2% of studies. 83% of studies took place in areas where non-consumptive (photographic) tourism occurred, 33% in an area with livestock ranching, game ranching, and/or farming, 15% in an area with trophy hunting, and 3% in a logging or mining concession. 71% of studies included a national of the study country as an author. For studies outside of South Africa, only 59% of studies included a national of the study country as an author. See Appendix S2 for a complete list of all studies and associated information, and Appendix S3 for additional details on the data interpretation.

## Taxonomic trends and biases

The lion was the species with the greatest number of population density assessments in peer-reviewed literature (55 studies which fitted the described criteria, leading to 90 estimates of population density), followed by spotted hyaena (34 studies, 81 estimates), leopard (33 studies, 71



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estimates), cheetah (19 studies, 21 estimates), brown hyaena (11 studies, 39 estimates), African wild dog (6 studies, 7 estimates), and finally striped hyaena (3 studies, 3 estimates) (Fig. 3). Results from the GLMM Model 1 (Fig. 4, Table S9) confirmed that more studies than average were conducted on lion (which exhibited a significant positive bias) and on spotted hyaena, leopard, brown hyaena and cheetah, and fewer than average on wild dog and striped hyaena (with the latter exhibiting a significant negative bias). Lion was also the species with the greatest study density (studies per 100,000 km² of range), with this being an order of magnitude greater than for all other species (Fig. 3).

After controlling for differences in geographical range (GLMM Model 2), there were still significantly fewer studies than expected on striped hyaena; in fact, the species' very large geographical range (Table S1) resulted in the negative bias for the species increasing (Table 1, Fig. 4). Lion still experienced a significant positive bias, while results suggest that the negative effect for African wild dog and the positive effect for spotted hyaena were partly explained by their restricted and large geographical ranges, respectively (Table S1; Fig. 4).

# Geographical trends and biases

- 239 Regional
- 240 Across all species, southern Africa was the region with the greatest number and density of large
- carnivore population assessments, followed by eastern Africa (Table 2).
- 242 The greatest number of published lion population assessments came from eastern Africa, followed
- by southern, central, and finally western Africa (see Table S1 for regional-level species-specific
- 244 insights). Nevertheless, western Africa exhibited the highest density of lion assessments, due to
- 245 the species' limited range in the region. Most leopard population assessments were carried out in



southern Africa, with more than twice the number of studies than in all other regions combined. Cheetah population assessments also primarily took place in southern Africa, followed by eastern (which exhibited the greatest study density) and northern Africa, with none in western or central Africa. African wild dog assessments were relatively evenly spread across the continent. Most spotted hyaena population estimates were from southern Africa (also exhibiting the greatest study density), followed by eastern and central Africa; no studies took place in western or northern Africa. All the only three published striped hyaena population assessments took place in eastern Africa. Finally, brown hyaena assessments were only carried out in Southern Africa, the only region where the species is present.

# Country

Across all species, geographical range had a significant positive effect (GLMM Model 2; Table 1), indicating that having more species range in a country was associated with more studies. For individual species (GLMs), geographical range had a significant positive effect on the number of lion, leopard, and spotted hyaena population assessments within a country, but not on those of cheetah (Table 4, Tables S10-S12).

South Africa was the country with the greatest number of large carnivore population assessments, followed by Tanzania, Kenya and Botswana. Study density, instead, was highest in Niger, followed by South Africa, Kenya, and Cameroon. Twenty-seven countries with at least one large carnivore species (57%) had no published density estimation studies (Table 3). Results from the GLMM Model 1 confirmed these results (Table S9). When controlling for range (GLMM Model 2), South Africa then Kenya still exhibited the strongest bias, while Tanzania dropped to fifth, with Cameroon and Botswana ranking higher (Table 1; Fig. 5). Chad, South Sudan, DRC, and Angola



5). 269 270 For individual species, Tanzania was the country with the greatest number of lion assessments, 271 followed by Kenya and South Africa. Uganda was the country with the highest lion study density; if we only consider countries with >40,000 km<sup>2</sup> of lion range, however, South Africa exhibited the 272 273 greatest study density, followed by Kenya, Zimbabwe, and Zambia. Central African Republic (hereafter CAR), Ethiopia, Namibia, and Mozambique all exhibited large species range, but few 274 studies. Six countries (24%) had no published lion population assessments (Table S2). The GLM 275 results revealed that Kenya, South Africa, and Cameroon had significantly more lion population 276 assessments than would be expected based on country range, while CAR and South Sudan the 277 fewest (Table 4). 278 For leopard, South Africa was the country with the most assessments by a wide margin (almost 279 half of all studies). Only 28% of leopard range states had a published estimate, and the two with 280 the greatest leopard range in Africa, Angola and the DRC, had none (Table S3). When controlling 281 for geographical range in the GLM, South Africa was an outlier, exhibiting the largest positive 282 283 bias in studies, and Zimbabwe, Cameroon and Kenya also had significantly more assessments than would be expected based on geographical range. Angola, DRC, and Ethiopia were the most 284 understudied (Table S10). 285 286 Botswana and Kenya were the countries with the most cheetah assessments, followed by Namibia and Tanzania. 60% of cheetah range states had no published assessments, including some with 287 large tracts of range (Chad, Ethiopia; Table S4). Kenya, Botswana, and Tanzania were the 288 countries with the strongest positive bias after controlling for range in the GLM, while Algeria, 289 290 Chad, and Ethiopia those exhibiting the strongest negative research bias (Table S11). Peer-

ranked lowest, due to the combination of large country range and a lack of studies (Table 3; Fig.



reviewed African wild dog population assessments were only carried out in four countries: South 291 Africa, Kenya, Zimbabwe, and CAR. Study density was highest in South Africa. Overall, 78% of 292 293 range countries had no published African wild dog population assessments, including the five countries with the largest geographical range (Table S5). 294 South Africa, Cameroon, Ethiopia, Kenya, and Namibia had the most spotted hyaena studies. 295 296 Cameroon had the highest study density, followed by South Africa and Congo. 24 range countries (63%) had no studies (Table S6), including Angola and DRC, the countries with the first and third 297 298 greatest species range. Cameroon, South Africa, and Namibia exhibited the largest values of the random intercepts when accounting for geographical range in the GLM, and Angola, DRC, and 299 300 Nigeria the smallest (Table S12). For striped hyaena, peer-reviewed assessments have only taken place in Kenya, with 96% of range states having no published density estimates (Table S7). Finally, 301 South Africa then Botswana were the countries with the most brown hyaena population 302 303 assessments, and Botswana, Zimbabwe, and South Africa those with the highest study density. 304 Four range countries (50%) had no studies (Table S8).

# Discussion

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# Taxonomic representativeness and biases

Over the past two decades, lion is the species that has received the greatest number of population assessments, both overall (Fig. 2) and relative to species' geographical range (Table 1). It is likely that the lion's highly charismatic nature (Macdonald et al., 2015), its role as a keystone and flagship species (Bauer et al., 2015b; Leader-Williams and Dublin, 2000), and its ability to attract conservation funding and generate revenue through tourism (Maciejewski and Kerley, 2014) all play a role. In addition, a relatively varied range of methodologies have been employed to survey



lion populations, including some ch cannot be typically be applied to most other large carnivore 313 species due to behavioural differences (e.g. 'call-ins'; Braczkowski et al., 2020). This likely also 314 315 played a part in the species receiving greater research attention. The large number of studies on spotted hyaena, the second most studied species, are instead likely 316 partly driven by its vast distribution (as indicated by our findings; Fig. 4), and by the relatively 317 318 large number of methods used to survey the species (Davis et al., 2022). In addition, assessments primarily targeting lions have often employed the data collected to also estimate spotted hyaena 319 320 density (e.g. Ferreira & Funston, 2016). Nevertheless, the fact that – due to the species' wide distribution – study density was still relatively low (Fig. 3) suggests that considerable knowledge 321 322 gaps remain across its range. 323 Striped hyaena had the lowest number of population assessments, and the strongest negative research bias (Table 1, Fig. 4). This is in line with previous suggestions that knowledge of the 324 species is particularly low (AbiSaid and Dloniak, 2015), and is likely due to a combination of 325 factors, including: populations often existing outside of formally protected areas (PAs; AbiSaid & 326 Dloniak, 2015); the species being more secretive and less well-known (Macdonald et al., 2015) 327 328 and thus less apt at raising conservation and tourism revenue (Di Minin et al., 2013; Okello et al., 2008); and the fact that much of its range is in northern Africa, a region exhibiting low levels of 329 conservation research in general (Agha et al., 2018; Hickisch et al., 2019; Trimble and van Aarde, 330 331 2012). Finally, the fact that some easily implementable methods used to survey other species (e.g. call-ins) cannot be reliably applied to striped hyaena (AbiSaid and Dloniak, 2015) is likely to also 332 have played a part. 333 334 The other species that exhibited a negative bias in research was wild dog (although this effect was not significant; Table 1, Fig. 4). This is of particular concern due to the species being the most 335



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threatened of the African large carnivores (Woodroffe and Sillero-Zubiri, 2020), and supports suggestions that conservation research can be poorly aligned with conservation priorities (Di Marco et al., 2017; Wilson et al., 2016). The observed paucity of estimates for wild dog is likely a result of the species being less well-known compared to the felids (Macdonald et al., 2015), and a combination of ecological and methodological factors. Wild dogs are social, low density species, with very large home ranges that often range outside of PA boundaries (Creel and Creel, 2002); as a result, survey methods often employed for other species (e.g. camera trapping combined with capture-recapture modelling; Strampelli et al., 2022) are challenging to apply to wild dogs, while others (e.g. call-ins) are less suitable due to behavioural characteristics. Indeed, although rapid assessment methods such as spoor counts have occasionally been employed to survey populations (Henschel et al., 2020), approximately half of published density estimates were obtained through resource-intensive long-term studies or citizen-science approaches, highlighting the difficulty associated with surveying the species rapidly. In addition, most wild dog populations in South Africa are part of an intensively managed metapopulation (Nicholson et al., 2020) where the exact number of individuals is known, and were therefore not eligible for our review, even though the populations are actively monitored. Leopard exhibited a slight positive bias in research, while there was little evidence of cheetah experiencing strong biases in either direction. The suitability of leopard to camera trap (Searle et al., 2021) and sign-based (Henschel et al., 2020) methods has facilitated assessments, as has its wide range and charismatic nature (Maciejewski and Kerley, 2014). On the other hand, although cheetahs are highly charismatic (Di Minin et al., 2013; Macdonald et al., 2015), listed as vulnerable to extinction (Durant et al., 2015), and has high potential to generate conservation revenue through tourism (Maciejewski and Kerley, 2014; Okello et al., 2008), a lack of rapid survey techniques for



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the species (Strampelli et al., 2021) likely played a role in survey efforts not being greater. 359 Additionally, the fact that – as for wild dog – numerous populations in South Africa inhabit small, 360 361 intensively managed reserves (Buk et al., 2018) also likely played a part in availability of estimates being relatively low. 362 For all species except cheetah, greater geographical range in a country was found to correspond to 363 364 greater research effort in that country (Table 4, Tables S10-12). Nevertheless, our findings suggest that a range of additional factors, including the species' ecology, charisma, ability to generate 365 366 conservation funding and tourism revenues, and applicability of different survey methods all play a role in driving the extent of population assessments for a species.

# Regional trends and biases

ater research effort in southern Africa (Table 2) is likely a result of greater conservation investments the region (Brockington and Scholfield, 2010; Wilson et al., 2016), as well as the favourable socio-economic characteristics of some countries (UNDP, 2020). For most species, eastern Africa was the second most studied region; this is likely due to the region still harbouring numerous important large carnivore populations (Table S1), as well as the long history of conservation investments and research in some countries (e.g. Kenya, Tanzania; Brockington & Scholfield, 2010). Low research effort in northern Africa is likely due to most large carnivore populations in the region existing at low density and outside of PAs (Belbachir et al., 2015), to conservation research investments being comparatively low, and to the instability of some countries (Di Marco et al., 2017; Wilson et al., 2016). Similarly, the low level of research in northern, western, and central Africa for most species (Table S1) mirrors the region's under-representation across wider



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conservation research (Hickisch et al., 2019; Trimble and van Aarde, 2012; Wilson et al., 2016). For central Africa, this negative bias kely to primarily be a consequence of high volatility and low conservation investments in the region (Brockington and Scholfield, 2010; FFP, 2020). In western Africa, the fact that conservation investments and wildlife-oriented tourism have historically been lower than in southern and eastern Africa (Brockington and Scholfield, 2010; UNWTO, 2018) has likely had an impact on research effort. Our findings largely agree with suggestions of biases towards carnivore research and conservation activities in eastern and southern Africa (Ray et al., 2005). Nevertheless, when accounting for greater geographical range in these region, the density of population assessments was actually lower than in central and western Africa for a number of species (Table S1). In addition, for several species, many assessments in eastern Africa are from a small number of populations; for example, approximately half (48%) of lion, and 75% of cheetah assessments in the region were from a single population, in the Serengeti-Mara (Appendix S2), even though this comprises only ~7% of lion's and ~8% of cheetah's eastern African range. Without studies from this one population, eastern Africa would exhibit the lowest lion study density of any range region. Tele findings thus suggest that – after accounting for differences in geographical range and the repeated sampling of a few populations – the observed biases in population assessments towards certain regions are less clearcut than they may initially appear, and that understudied populations remain even the regions receiving the most research attention.

# Country trends and biases

As lack of research can impede biodiversity conservation (Di Marco et al., 2017), the fact that 57% of African countries with large carnivore range did not have a single published, peer-reviewed population assessment is of concern. When accounting for differences in geographical range,



Angola, South Sudan, DRC, and Chad exhibited the strongest negative bias in research (Table 1), 404 likely primarily as a consequence of years of unrest and insecurity in these countries (FFP, 2020). 405 406 South Africa showed the strongest positive bias in published research (Table 1), with the greatest number of assessments for several species (Tables S2-S8). Conservation investments are 407 particularly high in South Africa (Brockington and Scholfield, 2010), likely explaining this bias. 408 409 Although Tanzania was the country with the second highest number of assessments, accounting for geographical range caused the country to exhibit only the fifth strongest positive bias (Table 410 411 1), indicating that part of this effect is due to the country being home to considerable large carnivore range. Furthermore, 60% of all published lion population assessments in the country 412 413 were from a single ecosystem (Serengeti), comprising only 8% of lion country range. A similar bias was also observed in Kenya, with 45% of lion studies being from the Mara system, even 414 though this accounted for only ~3\% of national lion range. These biases towards a small number 415 of well-studied populations suggest that the total number of studies may not be a representative 416 417 indicator of how well a country's large carnivore populations have been studied. Overall, our results suggest that the extent of published biological research in a country is 418 419 dependent on a range of factors, including socio-economic status, research history and interests of individuals and organisations, priorities of funding agencies and governments, in-country support 420 and capacity, and language barriers to publication (Griffiths and Dos Santos, 2012; Trimble and 421 422 van Aarde, 2012).

# Opportunities and recommendations

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Reducing the identified geographical and taxonomic biases in population assessments would help ensure that all species and areas of conservation importance have an adequate knowledge base



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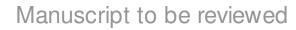
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available, improving their conservation outlook (Wilson et al., 2016). We argue that addressing the identified geographical biases (both by region, countries, and land use types) should be the most pressing priority; while a focus towards certain species, such as lion, may indirectly lead to conservation benefits for other species (e.g. by identifying areas requiring additional conservation investments), research from well-studied areas is unlikely to be able to help inform conservation decisions in poorly-studied regions (Di Marco et al., 2017). Thus, much like for the wider conservation research field (Trimble and van Aarde, 2012), geographical biases in research and assessments are more immediate hurdles for science-based conservation management of African large carnivores. As a result, northern, western, and central Africa should be considered priority regions for future research. Increased attention should in particular be given to the twenty-six countries which currently lack any published estimates (Table 3), especially Angola, DRC, South Sudan, and Chad, given their considerable large carnivore country ranges and their potential importance for the conservation of these species (Dickman et al., 2015). Within southern and eastern Africa, we encourage prioritising additional research outside of South Africa and the Serengeti-Mara ecosystem, respectively, to complement the efforts there. Shortfalls in conservation research in countries with high levels of biodiversity but low income can be due to a lack of trained and/or funded biologists (Gaston, 2000), research infrastructure (Wilson et al., 2016), or dedicated finances (Githiru et al., 2015). As a result, building capacity of researchers and practitioners in large carnivore survey and monitoring techniques in underrepresented areas should be a priority. The fact that only 59% of studies outside of South Africa included a co-author from the study country reinforces suggestions that research in developing countries is disproportionately led by scientists from more developed areas (Stocks et al., 2008), and shows there is considerable need for such capacity building efforts. For these reasons, we



recommend that both donors and foreign researchers maximise the involvement of local scientists, 449 students, and practitioners in future assessments, including through capacity building initiatives 450 such as the provision of training and funding or equipment. 451 Conservation donors and funders should encourage efforts in understudied regions, as well as for 452 understudied species, to ensure that conservation research occurs where it is most needed (Di 453 454 Marco et al., 2017; Wilson et al., 2016). This is especially important the case given widespread funding shortfalls in conservation (Lindsey et al., 2018). At the same time, we also urge funders 455 and practitioners to recognise the importance of scale dependency, as even within more studied 456 countries gaps remain: in Tanzania, the country with the greatest lion range and the most lion 457 studies, most populations are nevertheless yet to be assessed. We therefore emphasize the 458 importance of decision and investment processes being multi-scale, and of appreciating the 459 intricacies of the identified biases. 460 On a species level, we echo calls for further population assessments of striped hyaena (AbiSaid 461 and Dloniak, 2015). We also strongly recommend prioritising further population assessments of 462 wild dog, particularly due to the species' classification as 'Endangered' (Woodroffe and Sillero-463 Zubiri, 2020). Such efforts are especially required in countries that have been identified as critical 464 for the species, but where no recent assessments have been carried out (e.g. Botswana and 465 Tanzania; Kuiper et al., 2018). As the lack of well-established methods to rapidly survey wild dog 466 populations is a key reason for these knowledge gaps (Woodroffe and Sillero-Zubiri, 2020), 467 longer-term intensive monitoring studies such as those carried out prior to this review's considered 468 period (e.g. Creel & Creel, 1996), alongside further development of novel, cost-effective 469 470 methodologies (e.g. citizen-science techniques; Marnewick et al., 2014), are strongly





471	recommended. The possibility of monitoring populations through alternative status parameters,
472	such as species' occupancy (Henschel et al., 2020), should also be explored.
473	Our findings also highlight the urgent need for additional cheetah population assessments,
474	particularly in northern, western, and central Africa. Due to their large country ranges, studies in
475	Chad and Ethiopia should especially be considered a priority. As in the case of wild dog, we also
476	recommend further development and standardisation of cheetah population monitoring techniques,
477	including the exploration of novel citizen-science based approaches (Marnewick et al., 2014;
478	Weise et al., 2017).
479	For leopard, surveys are particularly recommended in the 72% of range states without published
480	estimates, particularly those with large geographical range (Angola, DRC, Ethiopia, and South
481	Sudan; Table S3). Although considered a highly adaptable species, leopard populations are
482	increasingly under threat (Stein et al., 2020). It is therefore important that research and monitoring
483	is carried out across the species' range, rather than in localised 'hubs', as is currently the case
484	(South Africa; Table S3).
485	For spotted hyaena, future efforts should prioritise populations in northern and western Africa, as
486	no assessments are available from these regions, as well as in countries with extensive range but
487	no surveys (Angola, DRC, Nigeria, Somalia, and South Sudan). For brown hyaena, we recommend
488	the prioritisation of assessments across different habitats and land-use types (Kent and Hill, 2013),
489	and in countries where the species is yet to be surveyed (Angola, Mozambique, Eswatini; Table
490	S8).
491	Study density for lion was lowest in central Africa, which should therefore be considered a priority
492	region for further assessments. Nevertheless, considerable opportunity for further work exists in



most range countries; even in Tanzania and Kenya, the two with the greatest number of assessments, most populations have not been empirically assessed. Overall, the fact that even for lion, the species with the strongest positive bias in research, considerable gaps still exist highlights the extent to which African large carnivore populations are presently understudied and undermonitored.

Finally, efforts should be made to address the observed sampling bias towards photographic tourism areas, and against trophy hunting areas. Trophy hunting areas in Africa cover a greater area than National Parks (Lindsey et al., 2007), and, given that accurate estimates of population size are crucial to ensure the sustainability of trophy hunting (Mweetwa et al., 2018), we recommend that future efforts attempt to bridge this gap. Similarly, although working on public and/or unprotected land can be logistically challenging (Agha et al., 2018), efforts should also be made to address the negative sampling bias associated with these areas, as our findings support the global pattern of biodiversity monitoring largely taking place within PA networks (Martin et al., 2012). The fact that some species still heavily occupy areas outside PAs (e.g. spotted hyaena, Bohm & Höner, 2015; cheetah, Weise et al., 2017), and that non-protected areas encompass the majority of wildlife habitat across the continent (Agha et al., 2018), highlights the need for future efforts to include boundary and non-protected areas.

# **Conclusions**

We carried out the first review of peer-reviewed African large carnivore population assessments, focusing on the last two decades, and empirically tested for geographical and taxonomic biases in effort. We found research biases towards lion and against striped hyaena, and to a lesser extent African wild dog. Assessments were biased towards southern and eastern Africa, while Northern,



western, and central Africa were generally under-studied. Non-protected and trophy hunting
areas were under-sampled compared to photographic tourism areas, and significant opportunities
exist for greater inclusion of host country national in such studies. Overall, we recommend the
biases we have identified are employed by researchers, practitioners, and policymakers to
address knowledge gaps and help inform future research and monitoring efforts.

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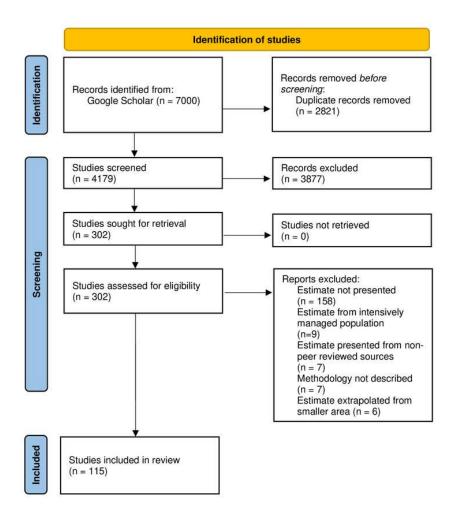
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PRISMA flow diagram

PRISMA flow diagram for methodology for syntethic review of African large carnivore assessments (2000-2020).







### Figure 2(on next page)

Location of peer-reviewed African large carnivore studies (2000-2020)

Location of peer-reviewed African large carnivore population assessment studies (2000 – 2020). Red crosses show estimates of population density, except in the case of brown hyaena, where blue crosses are used. When a study estimated multiple population densities, these are shown as separate points. Grey areas represent current geographical ranges (IUCN, 2020), except in the case of brown hyaena, where a dotted pattern is used.

**PeerJ** Manuscript to be reviewed Leopard Lion Panthera leo Panthera pardus ATT African wild dog Cheetah Lycaon pictus Acinonyx jubatus शर Striped hyaena Hyaena hyaena Spotted hyaena Crocuta crocuta Brown hyaena Parahyaena brunnea ATT

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**Figure 2.** Location of peer-reviewed African large carnivore population assessment studies (2000 – 2020). Red crosses show estimates of population density, except in the case of brown hyaena, where blue crosses are used. When a study estimated multiple population densities, these are shown as separate points. Grey areas represent current geographical ranges (IUCN, 2020), except in the case of brown hyaena, where a dotted pattern is used.



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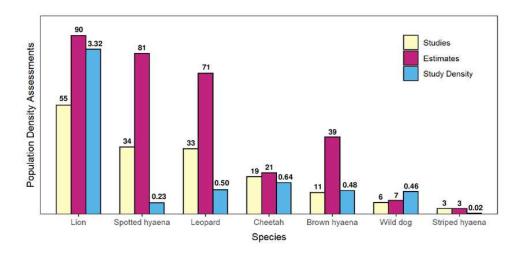
Summary of peer-reviewed African large carnivore density assessments, by species

Number of peer-reviewed population density assessments (studies), number of individual

population density estimates, and density of peer-reviewed studies (studies per 100,000 km²

of geographical range) for African large carnivores (2000 – 2020).



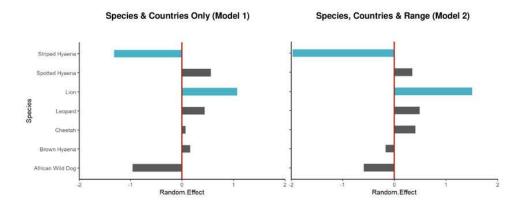


**Figure 3.** Number of peer-reviewed population density assessments (studies), number of individual population density estimates, and density of peer-reviewed studies (studies per  $100,000 \text{ km}^2$  of geographical range) for African large carnivores (2000 - 2020).

Random intercept values for species from the Poisson GLMMs

Random intercept values for species from the Poisson GLMM without explanatory variables (Model 1, Table S9) and for that accounting for geographical range (Model 2, Table 1). In both cases, values for lion and striped hyaena were credibly different from zero (i.e. zero not contained within the 95% HDI of the posterior distribution; shown in blue), with this being strongest when accounting for geographical range of species. The moderate negative effect for African wild dog and positive effect for spotted hyaena, on the other hand, were partly explained by their restricted and very large geographical ranges, respectively.





**Figure 4.** Random intercept values for species from the Poisson GLMM without explanatory variables (Model 1, Table S9) and for that accounting for geographical range (Model 2, Table 1). In both cases, values for lion and striped hyaena were credibly different from zero (i.e. zero not contained within the 95% HDI of the posterior distribution; shown in blue), with this being strongest when accounting for geographical range of species. The moderate negative effect for African wild dog and positive effect for spotted hyaena, on the other hand, were partly explained by their restricted and very large geographical ranges, respectively.



Number of large carnivore population assessment studies in Africa and random effects for individual countries

Left: Number of large carnivore population assessment studies in Africa, by country. Right: Random effects for individual countries from the Poisson GLMM investigating biases in large carnivore population assessments in Africa, accounting for differences in geographical range between countries (Model 2). A positive value (green) indicates more population assessments than expected based on large carnivore geographical range within the country, while a negative value (red) fewer (all species combined). See Table 1 for country specific values. For all figures in this study, country boundaries are based on the definitions of the African Union (https://web.archive.org/web/20130927110741/http://www.afrimap.org/english/images/report/AfriMAP-AU-Guide-EN.pdf



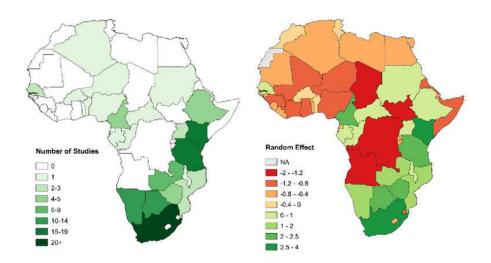


Figure 5. Left: Number of large carnivore population assessment studies in Africa, by country. Right: Random effects for individual countries from the Poisson GLMM investigating biases in large carnivore population assessments in Africa, accounting for differences in geographical range between countries (Model 2). A positive value (green) indicates more population assessments than expected based on large carnivore geographical range within the country, while a negative value (red) fewer (all species combined). See Table 1 for country specific values. For all figures in this study, country boundaries are based on the definitions of the African Union (https://web.archive.org/web/20130927110741/http://www.afrimap.org/english/images/report/AfriM AP-AU-Guide-EN.pdf).



#### Table 1(on next page)

Results of the Poisson GLMM investigating biases in large carnivore population assessments in Africa, accounting for differences in geographical range between countries (Model 2).

Results of the Poisson GLMM investigating biases in large carnivore population assessments in Africa, accounting for differences in geographical range between countries (Model 2). Species and countries are modelled as random effects. For species, a positive value indicates more assessment than expected given their geographical range, while a negative fewer. For countries, a positive value indicates more assessments than expected based on large carnivore geographical range within that country, while a negative fewer. Values credibly different from zero (i.e. 95% highest density interval (HDI) of the posterior distribution does not contain zero) are highlighted in bold for species and range.



**Table 1.** Results of the Poisson GLMM investigating biases in large carnivore population assessments in Africa, accounting for differences in geographical range between countries (Model 2). Species and countries are modelled as random effects. For species, a positive value indicates more assessment than expected given their geographical range, while a negative fewer. For countries, a positive value indicates more assessments than expected based on large carnivore geographical range within that country, while a negative fewer. Values credibly different from zero (i.e. 95% highest density interval (HDI) of the posterior distribution does not contain zero) are highlighted in bold for species and range.

Name	Mean	HDI low	HDI high	SD
Intercept	-2.07	-3.60	-0.65	0.75
Range *	0.60	0.26	0.93	0.17
Country Random Effect Variance	1.70	1.07	2.46	0.37
Species Random Effect Variance	1.46	0.52	2.78	0.69
Brown Hyaena	-0.17	-1.53	1.20	0.69
Cheetah	0.41	-0.91	1.75	0.67
Leopard	0.49	-0.76	1.80	0.65
Lion *	1.51	0.29	2.88	0.66
Spotted Hyaena	0.35	-0.93	1.65	0.65
Striped Hyaena †	-1.97	-3.82	-0.39	0.89
African Wild Dog	-0.59	-2.00	0.82	0.71
South Africa *	3.13	2.24	4.08	0.48
Kenya *	2.91	2.02	3.87	0.48
Cameroon *	2.47	1.45	3.56	0.54
Botswana *	2.38	1.43	3.38	0.50
Tanzania *	2.19	1.21	3.24	0.52
Zimbabwe *	2.10	1.05	3.20	0.55
Namibia *	1.68	0.62	2.78	0.55
Zambia *	1.48	0.36	2.64	0.58
Mozambique	1.04	-0.18	2.28	0.63
Uganda	0.96	-0.48	2.37	0.73
CAR	0.92	-0.40	2.20	0.66
Ethiopia	0.86	-0.38	2.12	0.63
Malawi	0.73	-0.95	2.31	0.83
Senegal	0.61	-1.04	2.18	0.82
Gabon	0.50	-1.59	2.48	1.04
Congo	0.44	-1.61	2.41	1.03
Nigeria	0.41	-1.71	2.34	1.03
Sudan	0.27	-1.34	1.79	0.79
Tunisia †	-0.16	-3.42	3.03	1.64
Benin †	-0.17	-2.04	1.68	0.95
Burkina Faso †	-0.19	-2.09	1.63	0.95
Morocco †	-0.36	-3.40	2.64	1.55
Algeria †	-0.44	-2.49	1.50	1.02
The Gambia †	-0.52	-3.48	2.29	1.48
Lesotho †	-0.55	-3.48	2.23	1.47
Equatorial Guinea †	-0.56	-3.37	2.28	1.46
Djibouti †	-0.58	-3.50	2.10	1.44



Egypt †	-0.70	-3.57	1.92	1.42
Mauritania †	-0.70	-3.51	1.96	1.42
Libya †	-0.74	-3.61	2.00	1.44
Rwanda †	-0.76	-3.56	1.86	1.40
Burundi †	-0.76	-3.54	1.85	1.40
Sierra Leone †	-0.77	-3.62	1.77	1.39
Guinea-Bissau †	-0.77	-3.56	1.82	1.39
Liberia †	-0.77	-3.59	1.81	1.39
Togo †	-0.80	-3.53	1.83	1.39
Eritrea †	-0.82	-3.54	1.78	1.38
Ghana †	-0.82	-3.57	1.72	1.37
Guinea †	-0.84	-3.62	1.65	1.37
Cote d'Ivoire †	-0.85	-3.58	1.68	1.37
Eswatini †	-0.86	-3.65	1.67	1.37
Somalia †	-1.04	-3.67	1.38	1.31
Nigeria †	-1.13	-3.69	1.26	1.29
Mali <sup>†</sup>	-1.14	-3.74	1.26	1.30
Niger †	-1.14	-3.79	1.20	1.29
Chad †	-1.42	-3.91	0.85	1.23
South Sudan †	-1.42	-3.91	0.84	1.24
DRC †	-1.57	-4.05	0.59	1.21
Angola †	-1.72	-4.09	0.47	1.18

**Legend:** HDI = Highest Density Interval; SD = standard deviation; Range = country range, based on IUCN Red List polygons (IUCN, 2020); \* = value suggests significant positive effect; † = value suggests significant negative effect.

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### Table 2(on next page)

Regional geographical trends in large carnivore research in Africa (2000 – 2020), all species combined

Regional geographical trends in large carnivore research in Africa (2000 – 2020), all species combined. All figures were obtained by combining estimates of geographical range, studies/estimates, and density of studies/estimates for all large carnivore species. See Table S1 for species-specific insights.



**Table 2.** Regional geographical trends in large carnivore research in Africa (2000 - 2020), all species combined. All figures were obtained by combining estimates of geographical range, studies/estimates, and density of studies/estimates for all large carnivore species. See Table S1 for species-specific insights.

Region	Geographical Range (km²)*	Studies	<b>Density Estimates</b>	Studies / 100,000 km²	Estimates / 100,000 km²
Southern	12,114,919	62	176	0.51	1.45
Eastern	12,587,456	38	86	0.30	0.68
Central	6,116,580	7	40	0.11	0.65
Western	4,832,540	5	8	0.10	0.17
Northern	8,145,160	1	1	0.01	0.01

<sup>\*</sup> Based on IUCN Red List geographical range polygons (IUCN, 2020)

1



### **Table 3**(on next page)

Large carnivore population assessments in Africa (2000-2020) by country, all species combined.

Large carnivore population assessments in Africa (2000-2020) by country, all species combined. Figures were obtained by combining estimates of geographical range, studies/estimates, and density of studies/estimates for all large carnivore species. Refer to Tables S2 – 8 for species-specific insights.



**Table 3.** Large carnivore population assessments in Africa (2000-2020) by country, all species combined. Figures were obtained by combining estimates of geographical range, studies/estimates, and density of studies/estimates for all large carnivore species. Refer to Tables S2 - 8 for species-specific insights.

Country	Studies	Estimates	Geographical Range (km²) *	Studies / 100,000 km <sup>2</sup>	Estimates / 100,000 km <sup>2</sup>
South Africa	27	88	1,601,900	1.69	5.49
Tanzania	16	41	2,871,700	0.56	1.43
Kenya	15	31	1,735,300	0.86	1.79
Botswana	11	38	2,469,150	0.45	1.54
Namibia	10	16	2,388,600	0.42	0.67
Zambia	7	12	1,305,900	0.54	0.92
Cameroon	5	29	600,740	0.83	4.83
Zimbabwe	4	13	752,800	0.53	1.73
Ethiopia	4	5	2,884,486	0.14	0.17
Mozambique	3	4	1,413,300	0.21	0.28
Uganda	2	7	468,900	0.43	1.49
Senegal	2	2	349,800	0.57	0.57
CAR	1	10	1,147,040	0.09	0.87
Malawi	1	6	135,600	0.73	4.42
Gabon	1	3	255,600	0.39	1.17
Nigeria	1	2	1,012,000	0.10	0.20
Sudan	1	2	1,729,400	0.06	0.12
Niger	1	1	17,190	5.82	5.82
Benin	1	1	300,400	0.33	0.33
Congo	1	1	386,000	0.26	0.26
Burkina Faso	1	1	572,550	0.17	0.17
Algeria	1	1	3,089,900	0.03	0.03
Angola	0	0	2,009,200	0.00	0.00
Burundi	0	0	28,700	0.00	0.00
Chad	0	0	1,940,300	0.00	0.00
Cote d'Ivoire	0	0	273,400	0.00	0.00
Djibouti	0	0	24,800	0.00	0.00
DRC	0	0	1,751,000	0.00	0.00
Egypt	0	0	1,015,800	0.00	0.00
Equatorial Guinea	0	0	12,800	0.00	0.00
Eritrea	0	0	257,798	0.00	0.00
Eswatini	0	0	38,369	0.00	0.00
Ghana	0	0	209,100	0.00	0.00
Guinea	0	0	277,100	0.00	0.00
Guinea-Bissau	0	0	43,000	0.00	0.00
Lesotho	0	0	100	0.00	0.00
Liberia	0	0	41,200	0.00	0.00
Libya	0	0	1,758,000	0.00	0.00
Mali	0	0	1,602,600	0.00	0.00



Morocco	0	0	976,900	0.00	0.00
Rwanda	0	0	28,500	0.00	0.00
Sierra Leone	0	0	62,700	0.00	0.00
Somalia	0	0	1,314,700	0.00	0.00
South Sudan	0	0	1,271,872	0.00	0.00
The Gambia	0	0	11,300	0.00	0.00
Togo	0	0	54,600	0.00	0.00
Tunisia	0	0	163,600	0.00	0.00

<sup>\*</sup> Based on IUCN Red List geographical range polygons (IUCN, 2020)

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#### **Table 4**(on next page)

Results of the models investigating biases in lion population assessments in Africa, accounting for geographical range (Negative Binomial GLM).

Results of the models investigating biases in lion population assessments in Africa, accounting for geographical range (Negative Binomial GLM). For residuals of individual countries, a positive value indicates more assessments than expected after controlling for the relevant variables, a negative value fewer. Significant residual values (defined as zero not contained within the 95% highest density interval (HDI) of the posterior distribution) are highlighted. Names of fixed effects and of countries are in bold if the variable exhibits a significant (i.e. credible non-zero) effect. Similar results for other species are presented in Appendix S4.



**Table 4.** Results of the models investigating biases in lion population assessments in Africa, accounting for geographical range (Negative Binomial GLM). For residuals of individual countries, a positive value indicates more assessments than expected after controlling for the relevant variables, a negative value fewer. Significant residual values (defined as zero not contained within the 95% highest density interval (HDI) of the posterior distribution) are highlighted. Names of fixed effects and of countries are in bold if the variable exhibits a significant (i.e. credible non-zero) effect. Similar results for other species are presented in Appendix S4.

	Name		Mean	HDI low	HDI high
(Intercept)			0.73	0.29	1.17
Range *			0.51	0.08	0.95
Kenya *			3.46	1.59	5.36
South Africa *		2.35	0.94	3.	73
Cameroon *		2.05	0.78	3	36
Zimbabwe		0.66	-0.01	1	34
Zambia		0.62	-0.08	1	33
Uganda		0.36	-0.25	0.9	98
Botswana		0.35	-0.50	1.2	21
Tanzania		0.28	-1.09	2.	12
Nigeria		-0.29	-0.67	0.	10
Sudan		-0.3	-0.67	0.	10
Burkina Faso		-0.31	-0.68	0.0	08
Benin		-0.31	-0.68	0.0	07
Senegal		-0.33	-0.69	0.0	04
Mozambique †		-0.52	-0.96	-0.	08
Ethiopia †		-0.63	-0.95	-0.	31
Namibia †		-0.68	-1.02	-0.	35
Niger †		-0.91	-1.21	-0.	63
Chad †		-0.91	-1.21	-0.	62
Malawi †		-0.92	-1.22	-0.	63
Angola †		-0.92	-1.23	-0.	63
DRC †		-0.92	-1.23	-0.	64
South Sudan †		-0.93	-1.24	-0.	64
CAR †		-0.99	-1.57	-0.	49