Priority areas for vulture conservation in the Horn of Africa largely fall outside the protected area network

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Summary

Vulture populations are in severe decline across Africa and prioritization of geographic areas for
their conservation is urgently needed. To do so, we compiled three independent datasets on vulture
occurrence from road-surveys, GPS-tracking, and citizen science (eBird), and used maximum
entropy to build ensemble species distribution models (SDMs). We then identified spatial vulture
conservation priorities in Ethiopia, a stronghold for vultures in Africa, while accounting for
uncertainty in our predictions. We were able to build robust distribution models for five vulture
species across the entirety of Ethiopia, including three Critically Endangered, one Endangered, and
one Near Threatened species. We show that priorities occur in the highlands of Ethiopia, which
provide particularly important habitat for Bearded Gypaetus barbatus, Hooded Necrosyrtes mon-
achus, Rüppell’s Gyps ruppelli and White-backed Gyps africanus Vultures, as well as the lowlands
of north-eastern Ethiopia, which are particularly valuable for the Egyptian Vulture Neophron
percnopterus. One-third of the core distribution of the Egyptian Vulture was protected, followed by
the White-backed Vulture at one-sixth, and all other species at one-tenth. Overall, only about one-
fifth of vulture priority areas were protected. Given that there is limited protection of priority areas
and that vultures range widely, we argue that measures of broad spatial and legislative scope will be
necessary to address drivers of vulture declines, including poisoning, energy infrastructure, and
climate change, while considering the local social context and aiding sustainable development.

Keywords: conservation prioritization, Ethiopia, vulture safe zones, ecological niche modeling,
species distribution model (SDM)

Introduction

Current species extinction rates are an estimated 1,000 times over the background rate and are
expected to increase with growing anthropogenic pressures worldwide (Barnosky et al. 2011). Loss
of species and wildlife abundance is compromising ecological processes, reducing ecosystem ser-
vices, and directly affecting humans (Şekercioğlu 2010, Cadotte et al. 2011). Tropical biodiversity is
generally less studied than temperate biodiversity (Trimble and van Aarde 2012) and the tropics are
expected to experience the greatest biodiversity losses in the 21st century (Alroy 2017). Overcom-
ing this trend will require international investment and local community support (Ghosh-Harihar
et al. 2019). As biodiversity and threats are unevenly distributed, limited research and conservation
investments should be strategically targeted to maximize effectiveness (Brooks et al. 2006) and to
fill threat gaps (Joppa et al. 2016).

As scavengers, vultures constitute the most endangered functional guild of birds and their
populations are in severe decline worldwide (Buechley and Şekercioğlu 2016a,b). They are a top
conservation priority because they are highly threatened and provide critical ecological functions
by quickly consuming carrion, contributing to nutrient cycling and regulation of problematic
facultative scavengers (Ogada et al. 2012) and, potentially, controlling disease (Buechley and
Vultures face a range of threats, including most notably, poisoning, toxic veterinary drugs, loss of food availability, and collision and electrocution on energy infrastructure (Buechley and Şekercioglu 2016a, Botha et al. 2017, Plaza et al. 2019). Given their broad distributions, large individual home ranges, and long-distance nomadic and migratory movements, conservation of vultures is challenging (e.g. Runge et al. 2014, 2015). Prioritizing geographic areas for conservation actions is needed (Mukherjee et al. 2014, Santangeli et al. 2019a). Indeed, the recently published Multi-Species Action Plan to Conserve African-Eurasian Vultures (Vulture MSAP), which is based on extensive expert input, provides a roadmap for vulture conservation across Africa and Eurasia, highlighting the need to identify threats and prioritize conservation actions at different spatial scales (Botha et al. 2017).

To identify spatial conservation priorities, it is essential to know the distribution and habitat use of species. Species distribution models (SDMs) are an important tool used to identify priority habitats and to forecast anthropogenic effects on species (Guisan and Thuiller 2005, Aryal et al. 2016). Combined with information on protected areas and existing threats to species, SDMs can help identify conservation priorities (Ferraz et al. 2012, Evans et al. 2018). However, the accuracy of SDMs is limited by the availability of both occurrence (i.e. where a species occurs) and predictor variables (i.e. spatial layers relevant to model a species’ distribution) (Fletcher Jr et al. 2018), both of which can be limited in the Global South. Targeted survey data are costly and time consuming, and thus generally scarce. Bio-logging (e.g. GPS tracking) can be used to model species’ distributions (e.g. Coxen et al. 2017), but can suffer from small sample sizes. Citizen science data (e.g. eBird; Sullivan et al. 2014) are being increasingly used to fill information gaps and have been shown to be useful for modelling bird distributions and population trends (Horns et al. 2018, Fink et al. 2019), but have spatial and observer biases (Fletcher Jr et al. 2018, Horns et al. 2018, Fink et al. 2019, Neate-Clegg et al. 2020). Combining multiple data sources to model species’ distributions can help increase predictive accuracy and account for sampling biases (Fletcher Jr et al. 2018, Miller et al. 2019).

Here we used an SDM framework to identify spatial conservation priorities for vultures in Ethiopia, a high-priority country for vulture conservation (Santangeli et al. 2019a). Our aims were to: 1) model the distribution of each vulture species; 2) identify vulture conservation priority areas; and 3) assess how well protected vulture priority areas are. We incorporated three independent datasets on vulture occurrence (road-surveys, GPS-tracking, and citizen science) to build robust predictions on vulture space use. Our results provide a framework for where to prioritize conservation work on endangered vultures in the Horn of Africa.

Methods

Study location and species

The study took place in Ethiopia, a biologically diverse and little-studied country, which supports one of the most species rich and abundant vulture communities worldwide (Mundy et al. 1993, Buechley et al. 2019). It is also the second most populous country in Africa and has a high human population growth rate (2.46% per year) (World Bank 2019b). Despite being one of the poorest countries in the world, Ethiopia is developing rapidly (World Bank 2019a), and there is severe pressure on natural ecosystems. Seven vulture species reside in Ethiopia: Bearded Gypaetus barbatus (‘Near Threatened’), Egyptian Neophron percnopterus (‘Endangered’), Hooded Necrosyrtes monachus (‘Critically Endangered’), Lappet-faced Torgos tracheliotos (‘Endangered’), Rüppell’s Gyps rueppelli (‘Critically Endangered’), White-backed Gyps africanus (‘Critically Endangered’), and White-headed Trigonoceps occipitalis (‘Critically Endangered’) (Botha et al. 2017). The largest African populations of Bearded and Egyptian Vultures are thought to occur in Ethiopia (Arkumarev et al. 2014, Botha et al. 2017).
Species occurrence data

Vulture occurrence data were acquired from three independent sources: road-surveys, satellite telemetry, and citizen science. Road-surveys took place in Ethiopia from 2010 to 2018, whereby two experienced raptor biologists completed 10,857 km of surveys, while driving at speeds <60 km per hour during daylight hours, identifying perched and flying vultures within 1 km of the road. Citizen science observations of vultures were downloaded via the comprehensive eBird Basic Dataset (Sullivan et al. 2014) for Ethiopia, and censored to minimise sampling and observer bias, following specific recommendations and code provided by Johnston et al. (2019). Satellite tracking data were collected by deploying 34 solar-powered GPS-transmitters on vultures in or ranging through Ethiopia from 2012 to 2018, including 15 Egyptian, eight White-backed, six Hooded, three Ruppell’s, and two Lappet-faced vultures. See Figure S1 in the online supplementary material for an overview of occurrence data. Further details on acquisition and treatment of occurrence data are in Appendix S1.

Each dataset had strengths and weaknesses. For example, road-surveys were designed to have broad geographic coverage and high data quality, but were limited by the extent of the road network and security concerns inhibiting access to some areas. Further, road-surveys took place only during the dry season, and seasonal differences in vulture ranging behaviour has been shown elsewhere in Africa (e.g. Kendall et al. 2014), indicating that these surveys may not have captured each species’ full environmental niche. In contrast, satellite-tracking provided a picture of the habitat use of individual vultures throughout the full annual cycle, but had a limited sample of individuals. Meanwhile, citizen science data had broad geographic and temporal coverage, but were spatially biased towards urban centres and birding “hotspots”, and may have suffered from lower data quality due to potential species misidentification (e.g. Fink et al. 2019). We recognize that each of the data types was therefore capturing different aspects of the species’ ecological niche and at somewhat different spatial and temporal scales, and we see that as a net benefit of our approach. By incorporating multiple data types, we maximized the number of species for which we had sufficient data to model distributions, while also incorporating data that captured different aspects of each species’ environmental niche.

All analyses were completed in R (Version 3.1.0; R Core Team 2019). For eBird data, occurrence points were observations for each species spatially rarefied to a 1-km distance, while the background sample was the location of all complete checklists (Coxen et al. 2017). For road-survey data, we buffered road transects by 1 km on both sides of the road, overlaid a 1-km² raster grid over the survey area, and aggregated observations for each species within each pixel, such that any pixel where a given vulture species was sighted was given a value of 1 (occurrence), whereas pixels where that vulture species was not encountered were given a value of 0 (background). For satellite tracking data, we calculated the 95% minimum convex polygon (MCP) cumulative home range (‘adehabitatHR’ package; Clemente Calenge 2011) of all tracked individuals of each species (with one exception: we calculated separate MCPs for two Hooded Vulture populations that did not overlap from transmitter deployments in northern and southern Ethiopia). Tracked vulture locations within each species’ MCP were used as occurrence points, while background points were drawn from a systematic sample of 10,000 evenly spaced points within the MCP for each species (Benson 2013).

Environmental variables

We compiled environmental data that were expected to influence vulture use, including habitat type (European Space Agency 2017), elevation and ruggedness (Robinson et al. 2014), human footprint (Venter et al. 2016), climate (annually averaged temperature, precipitation, wind, solar radiation; Fick and Hijmans 2017), latitude and longitude. Prior to modelling, we assessed correlation of predictor variables by creating a correlation matrix using ‘corrplot’ (Wei and Simko 2013), and reviewed the variance inflation factors using the ‘usdm’ package (Naimi 2015). We used a
cutoff of $|r| = 0.60$ as an indication of strong co-linearity (Crandall et al. 2015). Elevation and temperature were inversely correlated ($|r| = -1$) and we thus excluded elevation because temperature was expected to have a more direct biological effect. Wind, precipitation, and longitude were highly correlated with other variables and were removed, leaving us with 12 predictor variables: latitude, ruggedness, human footprint, distance to cropland, distance to desert, distance to forest, distance to grassland, distance to shrubland, distance to urban, categorical landcover class, temperature, and solar radiation (Figure S3). Further details on processing of predictor variables can be found in Appendix S1.

Species distribution models

We modelled the distributions of seven vulture species in Ethiopia using maximum entropy (Maxent; Phillips et al. 2006). Maxent is the most widely used species distribution modeling technique (Elith et al. 2011). To maximize the accuracy of Maxent models, it is important to account for sampling bias of occurrence points, to select the appropriate background sample, and to select the best regularization multiplier (Merow et al. 2013). To address these concerns, we spatially thinned each dataset, carefully selected the background sample, and tested a range of regularization multipliers. Maxent was run using the package ‘ENMeval’ (Muscarella et al. 2014), using “algorithm = ‘maxent.jar’”, which interfaces with the standalone Maxent program (Version 3.4.1; Phillips and Dudik 2020). For each data type and species combination, we ran a suite of models with a range of regularization multiplier values (from 1 to 4, by 1) and feature class combinations (L, LQ, H, LQH, LQHP, LQHPT; where L = linear, Q = quadratic, H = hinge, P = product and T = threshold), or up to 24 models each. To assess the predictive capacity of each model, we used $k$-fold internal cross validation (Merow et al. 2013) with four folds (we used 75% of the data to train the model and 25% of the data to test the model), using the ‘block’ data partitioning method, which spatially partitions data into four bins by the lines of latitude and longitude that divide occurrence localities as equally as possible. We selected the top model for each species and data type using Akaike information criteria corrected for small sample sizes (AICc) (Burnham and Anderson 2004, Muscarella et al. 2014). Predictive accuracy of top models was assessed from the AUC$_\text{test}$ score, which measures the model’s ability to discriminate between conditions at occurrence versus background locations (Muscarella et al. 2014). We also report OR$_\text{MTP}$ which is the average omission rate of the occurrence records at the minimum training presence (MTP) threshold, where MTP represents an estimate of species habitat suitability (Muscarella et al. 2014, Taylor et al. 2020). We deemed models with AUC$_\text{test}$ values $>$0.60 to be informative (Randin et al. 2006).

Next, we compared the top predictions for each species and data type, using Hellinger’s $I$ statistic. Hellinger’s $I$ values range from 0 (no overlap) to 1 (complete overlap), quantifying niche overlap over geographic space while making no assumptions about species density (Warren et al. 2008).

Then for each species we created an ensemble species distribution model. To do so, we weighted the top species-data type model by its respective AUC test score and then summed them. We mapped the final predicted ensemble distribution of each species across Ethiopia, interpreted as relative habitat suitability (Merow et al. 2013).

To estimate uncertainty in our predictions, we took 10 bootstrap samples of the training datasets and ran predictions for each species and data type combination on each bootstrapped sample. We then calculated the standard deviation of the predicted values for each species and data type combination and derived an overall map of the uncertainty in the modelled distribution for each species by taking the average of the standard deviation values across each data type.

We identified the core distribution of each species within Ethiopia as the top 30% of its predicted distribution areas. While the 30% threshold is somewhat arbitrary, it follows other recent work prioritizing conservation areas for vultures across the Old World (Santangeli et al. 2019a). We then calculated overlap of the core distribution with protected areas (PAs; considering IUCN protected area categories I to VI; IUCN and UNEP-WCMC 2018) and Important Bird and Biodiversity Areas (IBAs; BirdLife International 2020).
Identifying priority conservation areas

We combined the final ensemble species distributions for each vulture species to identify priority conservation areas for all vultures in Ethiopia using software Zonation Version 4.0 (Moilanen et al. 2014). Zonation ranks areas of conservation importance by iteratively removing grid cells starting from those that have the lowest value for conservation. We used the core area cell removal method (CAZ), such that high priority areas include important habitat for each species, i.e. a ‘leave no species behind’ approach, at a 1-km² pixel resolution (Santangeli et al. 2019a). In order to assess the robustness of the main priority map, we incorporated the prediction uncertainty layers we created (see above) directly into the prioritization analysis using the distribution discounting tool in Zonation (Moilanen et al., 2006), such that the highest-ranking sites have high conservation value and low uncertainty. We ran the zonation analyses three times, setting the uncertainty parameter to subtract 0, 0.5, and 1 SD, from the nominal estimates of the input layers of analysis. We report on the intermediary output (0.5 SD) in the main text and provide the other outputs in Figure S5. Through the prioritization, all species were given the same weight. An alternative approach could be to weight species according to their global conservation status (e.g. Santangeli et al. 2019). However, we decided to give species equal weight, provided that our exercise was national, and considering that Ethiopia may provide particularly important habitat for some of the less threatened species, particularly Bearded and Egyptian Vultures (Botha et al. 2017). Next, we identified “vulture priority areas” in Ethiopia as the top 30% of the zonation output and calculated the area of overlap and the average priority of each PA and IBA in the country for vulture conservation.

Results

Species distribution models

We produced an ensemble SDM for five of the seven vulture species considered. Ensemble models were based on three data types for Egyptian and Hooded, two data types for Bearded and Rüppell’s, and one data type for White-backed Vulture (Table 1). AUCtest values for models used ranged from 0.606 for the Rüppell’s Vulture road-survey model to 0.928 for the Egyptian Vulture road-survey model (Table 1). Predicted distributions for each species from different data types showed a high level of niche overlap (Hellinger’s I statistic values ranged from 0.759 to 0.973; Table S1), indicating that the different datasets had largely similar predictions of each species’ distribution.

The leading predictor of Bearded Vulture occurrence was lower temperature, which was equivalent to higher elevations in Ethiopia, as well as more rugged areas. In contrast, Egyptian Vulture occurrence was best predicted by hotter temperatures, i.e. lowland desert areas. Hooded Vulture favoured human-dominated landscapes, while avoiding shrubland. Rüppell’s Vulture distribution was primarily predicted by ruggedness, as well as cooler, higher elevations. White-backed Vulture favoured cropland and forest habitats, proximity to urban centres, and cooler/ higher elevations.

The core range of the Egyptian Vulture was best protected (PA coverage = 32.8%, IBA coverage = 7.8%), followed by White-backed (PA = 14.3%, IBA = 5.2%), Rüppell’s (PA = 11.9%, IBA = 9.6%), Bearded (PA = 10.5%, IBA = 13.1%), and Hooded (PA = 9.3%, IBA = 7.1%) Vultures (Table S2). The final ensemble species distribution models for each species are shown in Figure 1 and the average variable importance in the final ensemble model for each species is shown in Figure 2.

Priority conservation areas

Vulture conservation priority areas occurred throughout the highlands of Ethiopia, as well as the lowlands in the north-east of the country (Figure 3a). Southern and eastern Ethiopia had generally
lower priority. The highest priority protected areas included Simien Mountains and Yangudi-Rassa National Parks, as well as several national forests (Figure 3b). These results were robust to uncertainty in predicted vulture distributions (Figure S5). Most of the highest priority IBAs were located in the vicinity of the capital city Addis Ababa in central Ethiopia (Figure 3c). The top 30% vulture priority areas encompassed an area of 338,388 km², of which 19.4% was covered by protected areas and 8.6% was covered by IBAs. See Appendices S2 and S3 for a full list of protected areas and IBAs with their priority rank.

### Table 1. Features and evaluation metrics of Maxent distribution models for seven vulture species in Ethiopia based on up to three data types. AUC_{test} measures the model's ability to discriminate between conditions at test versus background locations, while OR_{MTP} measures model overfitting (see text for details). For eBird and road-survey models, n refers to the number of observations, whereas for telemetry models n refers to the number of tracked individuals. Models with AUC_{test} < 0.6, shaded in grey, were deemed uninformative and excluded from further analyses. Models were not run for species lacking a sufficient sample size (<20 records for eBird or road-surveys, or <5 individuals for telemetry) and thus their values are NAs, but are included here to show sample size (n).

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Discussion

We provide the first quantitative analysis of vulture distributions in Ethiopia to identify spatial conservation priorities. We created ensemble models with up to three data sources per species to incorporate as much information as possible regarding each species' ecological niche. Priority areas occurred broadly in the central highlands of Ethiopia, which provide particularly important habitat for Bearded, Hooded, Rüppell's and White-backed Vultures, as well as the lowlands of northeastern Ethiopia, which are particularly valuable for the Egyptian Vulture (Figure 1, Figure 3). One-fifth of vulture priority areas were covered by protected areas, while less than one-tenth were covered by IBAs. Species core ranges were variably protected, with the Egyptian Vulture faring best at one-third protected, followed by White-backed Vulture at one-sixth, and all other species at one-tenth. That the vast majority (i.e. approximately 80%) of priority areas for vultures in Ethiopia fall outside the protected area network is in contrast to other studies in Africa that have shown vultures
persist primarily within protected areas across the continent (Thiollay 2006, Virani et al. 2011, Pomeroy et al. 2014, Murn et al. 2016), with exceptions for Egyptian and Hooded Vultures (Buechley et al. 2018, Henriques et al. 2018). Our results are strikingly similar to global Old World vulture conservation priority areas, for which one-fifth of the top-priority areas are protected and about one-tenth intersect IBAs (Santangeli et al. 2019a).

The limited protection of vulture priority areas indicates that conservation actions must be effective outside the existing protected area network. As has been argued at the global level (Botha et al. 2017, Santangeli et al. 2019a), measures of broad spatial and legislative scope are likely to be necessary to conserve vultures and their respective ecosystem services. Such measures could include legislation aimed at controlling the availability and use of veterinary drugs and poisons that are drivers of vulture declines across Africa (Ogada 2014). Further, comprehensive environmental impact assessments should occur prior to energy infrastructure developments to evaluate and mitigate threats to vultures and other biodiversity (Santangeli et al. 2019a). It will also be imperative to work closely and collaboratively with stakeholders to identify and address local issues, such as human-wildlife conflict, which could contribute to vulture declines. Given that Ethiopia has a disproportionate role in supporting vulture populations (Botha et al. 2017, Buechley et al. 2019, Santangeli et al. 2019a), yet has among the least economic resources globally (World Bank 2019a), we call for increased international attention to and funding for vulture conservation efforts here.

An important next step to improve vulture conservation in the Horn of Africa will be a more detailed assessment of the spatial distribution of threats (Botha et al. 2017). The use of poisons to
Figure 3. Spatial conservation priorities for vultures in Ethiopia. Panel a) shows the overall spatial priorities, which is the output of the zonation analysis. Panels b) and c) show the rank priority of the protected area (PAs) and Important Bird and Biodiversity Area (IBAs) networks, as well as the top-ten highest priority site. The colours of the filled polygons correspond with the scale, ranging from 0 to 1, which is a relative measure of spatial conservation importance for vultures.
control feral dog populations by municipalities is apparently widespread and a major danger for vulture populations (Abebe 2013). Further, electrocution on powerlines has been shown to kill Egyptian Vultures in neighboring Sudan (Angelov et al. 2012), and, similarly concerning levels of electrocution of vultures have been recently reported in Ethiopia (Bakari et al. 2020). Ethiopia also has one of the largest wind energy facilities in Africa, with ambitious development targets to grow this sector (Asress et al. 2013). While renewable energy expansion is generally a positive strategy for the development of Ethiopia and global environmental sustainability, poorly situated turbines and high voltage transmission lines can be highly detrimental to soaring birds and other wildlife (Barrios and Rodríguez 2004, Sánchez-Zapata et al. 2016). Climate change is also expected to severely impact the region (Soultan et al. 2019), and, given that temperature was one of the most important predictors for the occurrence of Bearded, Rüppell’s, and White-backed Vultures (Figure 2), it could be contributing to range contractions (Simmons and Jenkins 2007, Phipps et al. 2017). Further, we call for a review and update of the IBA network in Ethiopia, which is outdated and largely focused on wetland and riparian areas. The vulture conservation priority areas that were identified herein could be the basis for several additional IBAs, per the A1 Global IBA Criterion for globally threatened species (BirdLife International 2020).

The predicted distributions of vultures herein indicate that Egyptian, Hooded, and White-backed Vultures favour human dominated landscapes and proximity to urban areas. The Egyptian Vulture is known to regularly congregate at refuse dumps (Tauler-Ametller et al. 2017, Buechley et al. 2018, McGrady et al. 2019), and Hooded Vultures are known to associate with humans, particularly in West Africa and Ethiopia (Mundy et al. 1993, Mullié et al. 2017, Henriques et al. 2018, Thompson et al. 2020). In contrast, the White-backed Vulture is more widely regarded as averse to human dominated landscapes, or having already declined precipitously in them, to the point that it occurs mainly within protected areas across much of Africa (e.g. Thiollay 2006, Virani et al. 2011, Pomeroy et al. 2014; although note that tracked immatures in southern Africa spent the majority of their time outside of protected areas (Phipps et al. 2013). Provided that we tracked Hooded and Egyptian Vultures that were caught in the vicinity of towns, there could be some sampling bias in our models. However, these results are fully congruent with our extensive observations throughout the country. Vulture coexistence with humans in Ethiopia may be regarded as a mutualistic relationship (Gangoso et al. 2013, Moleón et al. 2014). Vulture ecosystem services, while rarely robustly quantified, are now widely recognized (Şekercioğlu 2006, Buechley and Şekercioğlu 2016a,b, Devault et al. 2016), including efficient nutrient cycling and carrion removal, reduction in greenhouse gas emissions (Morales-Reyes et al. 2015, 2017), controlling problematic facultative scavengers and insects (Ogada et al. 2012b, Buechley and Şekercioğlu 2016a), and potentially limiting the spread of disease (Markandya et al. 2008, Buechley and Şekercioğlu 2016a, Şekercioğlu et al. 2016, Plaza et al. 2020). Nonetheless, the association of some species of vultures with humans, and their reliance on human waste may constitute an ecological trap for vultures in the 21st century (Ogada et al. 2012a, Buechley and Şekercioğlu 2016a).

Notably, we found that the Egyptian Vulture distribution was concentrated in north-eastern Ethiopia. Other research has shown the importance of this area for the species, particularly for overwintering migrant populations (Arkumarev et al. 2014, Buechley et al. 2018). However, surveys and observations of the species in Ethiopia in the 20th century (Ash and Atkins 2009) indicated a more widespread distribution, with resident populations throughout the country. Our sample of occurrence records for this species was likely biased towards migrants because 13 of 15 tracked Egyptian Vultures were migrants and road surveys were done when most migrants had already arrived in Ethiopia. Nonetheless, that we had few records across much of Ethiopia indicates potential decline in the resident population in Ethiopia, and merits further investigation.

Unfortunately, we were unable to produce informative models of the distributions of Lappet-faced and White-headed Vultures, due to low sample sizes. Surveys in the last century indicated widespread distributions of both of these species in Ethiopia (Ash and Atkins 2009), and recent work has predicted occurrence of the latter species in Ethiopia (Murn et al. 2016). However, these species were the least reported from citizen science observations and road surveys (Table 1). This is
not unexpected, as these species are more solitary and averse to human activity. We encourage further targeted survey work on these two species to better understand their distributions, status, and priority conservation areas in the Horn of Africa. Additionally, SDMs built on telemetry and citizen science data from White backed Vulture were uninformative (AUC <0.6; Table 1). We believe this occurred because the White-backed Vulture is a generalist species that can be found across all habitats and elevations, and in both human-dominated and wild environments in Ethiopia, which made it difficult to identify strong signals of habitat association in our models.

There are several potential limitations to our data sources and analyses. In addition to the paucity of data for certain species, as discussed above, we had sparse data from the Somali Region of southeastern Ethiopia due to security issues inhibiting access. Thus, we caution that the modelled predictions for this region are extrapolatory and should be taken as a first indication of potential areas to target further investigations. Further, we recognize that spatial sampling biases across the three data types could have influenced our models. For example, road surveys were restricted to areas proximate to roads, citizen science data tend to be concentrated in human-populated areas, and tracking data may be influenced by individual behaviour and tagging locations. We worked to address these potential biases, following recommendations in Merow et al. (2013), by limiting the background sample in each species-data type model to areas that were surveyed or realistically accessible to tracked individuals. To assess how robust our predictions are, we also directly quantified niche overlap between different species-data-type model predictions (Table S1), quantified uncertainty in our predictions (Figure S4, Figure S5), and deducted areas with more uncertainty in our prioritization exercise. Overall results indicated that our models were largely robust to uncertainty. Further, our models lacked some predictor variables that could be important for vulture space use (for example, the location of carcasses). We expect that predictions could thus be further refined if predictive layers existed that more accurately captured important local aspects of vulture ecology (see relevant discussion here: Efrat et al. 2020, Santangeli et al. 2020).

In summary, we provide here the first detailed assessment of vulture distributions and spatial conservation priorities for one of the most important countries for vulture conservation worldwide. The low coverage of vulture priority areas by protected areas, as well as the association between vultures and human-dominated landscapes, indicates a need for actions of wide temporal and spatial scope (Botha et al. 2017, Santangeli et al. 2019a, Perrig et al. 2020). Enactment and enforcement of legislation will likely be essential to reduce the availability and use of veterinary drugs and poisons that threaten wildlife (Margalida et al. 2014, Ogada 2014). Similarly, use of best practices and thorough implementation of environmental impact assessments prior to energy development projects are necessary. Several international conventions and frameworks exist to facilitate development and enforcement of policy relevant to vulture conservation. For example, Ethiopia is a signatory of the Convention on Biological Diversity (CBD), Convention of Migratory Species (CMS), and the Convention on the International Trade of Endangered Species of Wild Fauna and Flora (CITES), which provide guidelines to support the conservation of wildlife, including vultures. Encouragingly, in 2020 Ethiopia also became a signatory of the Raptors MoU, under CMS, which supports specific actions to reduce mortality of raptors and vultures from poisoning and energy infrastructure (Botha et al. 2017). Such actions, if taken while considering the local social context and in a manner that aids sustainable development, would help to conserve threatened vultures and their critical ecological functions, while also benefiting many other species occurring in this biodiversity hotspot region of global importance. Further, we believe that it is imperative to reduce social and economic inequalities, promote peace, and invest in community-led initiatives in order to achieve biodiversity conservation in Ethiopia (Santangeli et al. 2019b), especially provided the recent increase in social unrest and violence in the country.

Supplementary Materials

To view supplementary material for this article, please visit http://doi.org/10.1017/S0959270921000228.
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Vulture conservation priorities in the Horn of Africa


