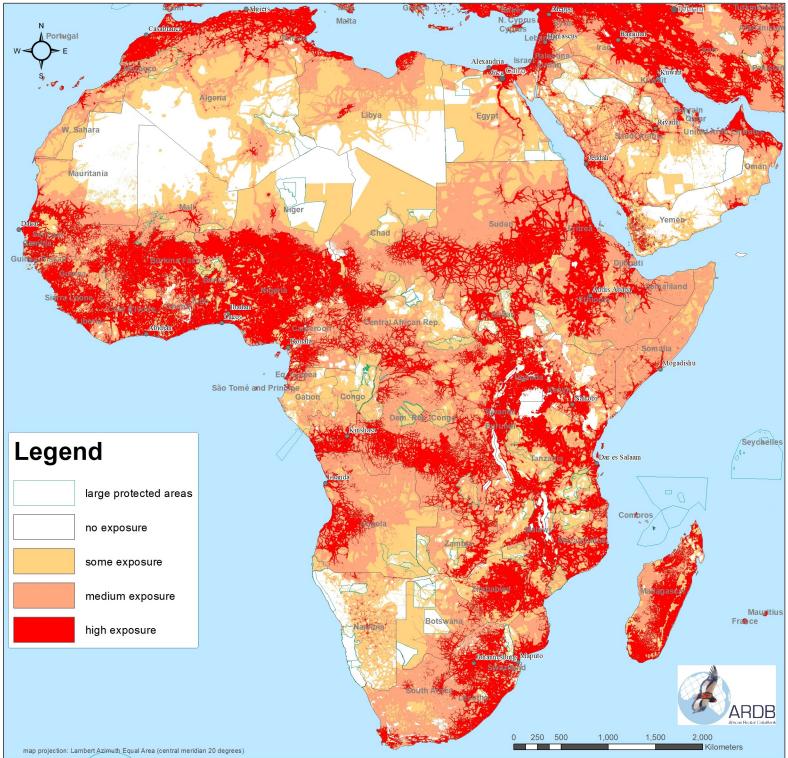
threat map: exposure to people

































METHODS

As before, this layer representing exposure to people is computed from three input layers: traveltime to cities (>50k), human population density and levels of protection derived as dollar spend per km2 within protected areas. Each of these layers can have extremely high maxima so we used log values of each to combine in order to show the subtle changes in rural areas across. Africa and Arabi. we were able to access new data for the entire modelled study area of Africa and Arabia, extending these data into the whole of the Arabian peninsula and nearby Asian countries (principally Turkmenistan). For traveltime to cities we tested new datasets developed for the World kby Habitat Info and also a more recent (2015) iteration by Nelson for traveltime to any urban areas but we found that the original traveltime data showed the best road network resolution. (http://biokupal.rc.ec.europe.eu/products/ggam/download.htm. - Nelson 2008). For human population density we update this layer to 2020 estimates from three sources: Gridded Population of the World v4 was used as a starter layer but was superseded by AfriPOP continental datasets for Africa and for Asia because of they are more finely modelled datasets. All three had been corrected to INDP population sizes for nations. We accessed the most recent data on protected areas from UNEP-WCMC (2023) and updated dollar spend per km2 within these protected areas from evious 1909 so values (James et al. 1.999) to 2012-2008 values using national biotevity spend data from Weldron et al. (2013) which include aid funding as well as domestic and other funding. We filled in missing values by literature search for Turkey, Introduced and from the three datasets were reclassified to a scale of 1-9 and combined as (traveltime velos of protection) – population density. In the present study we inverted the values and reclassified into the following four classes: 0 = no exposure threat, 1 = exposure threat present, 2 = medium exposure threat, 1 = exposure threat in the population de

CREDITS: Coordination: Andre Botha (EWT), Raigh Buij (Wageningen University & Research), Corinne Kendall (North Carolina Zoo), Ara Monadjem (University of Swaziland), Data collation: Lutfor Rahman & Lou Luddington (Habdat Info), Chris McClure, Evan Buschley, Leah Dunn (TFF). Analysis & map production: Leah Dunn (TFF) & Rob Daves (Habitat Info). Finance: The vulture surveys, data gathering and habitat and threat modeling were tended by the following organisations through Wageningen Diminently & Research (which also contributed resources): Dutch Ministry of Economic Alfans, WWF-Netherlands, UNFF-OMS Raptors MoU, North Carolina Zoo, Fondation LeFal Nature, Quagaja Foundation, Stiching Vorglepark Avriauna, Stiching Konnikijae Rotterdamse Dergaarde, Detrot Zoological Society, and Stiching Widlet. Through The Pereginner Furd this project benefied greatly from access to the ESRI Grant Scheme. Data on vultures were until but project benefied greatly from access to the ESRI Grant Scheme. Data on vultures were benefied, and with stimulation of the Stindard by the following individuals: Yilma D Abebe, Hichem Azatzal, Lalia Bahaa El Din, Nell & La Baker, Clive R Barlow, Keth Bildstein, Claire Bracebridge, Andy Branield, Erik & Asaph Brinaugh, Josof Browner, Chris Brothown, Evan Buschley, Raiph Buija & Barbara Croes, Andre Botha, Mike Cadman, Alazzar Daks Rulo, Rob Davies, Maria Disember, Waller Scheme, Peregory Katterocker, Adam Kane, Chris Kelly, Alan A Meg Kern, Conner Kendall, Holger & Galier Roberge Brantan, Alazzar Daks Ruly Madded, John Mendelsohn, Mike McGrady, Ara Monadjem, Campbell Murn, Ran Nathan, Kann Nelson, Sloyan Nikolov, Darcy Ogada, Stellen Opphel, Louis Phipps, Bram Piot, Thomas Rabell, Saachs Roberger, Standard, Holger & Galier Roberger, Branner, Brind, Madde, John Mandel, Short, Maria Roberger, Aller Roberger, Aller Roberger, Branner, Brind, Thomas Rabell, Saachs Roberger, Standard, Karper, Commer Kendall, Holger & Galier Roberger, Branner, Brind, Madde, John Nauden, Schrift Roberger, Aller Roberger, Al TRUST, ROYAL SOCIETY FOR THE PROTECTION OF BIRDS, TANZANIAN BIRD ATLAS, THE PEREGRINE FUND, UNIVERSITY OF UTAH, VULPRO, WEST AFRICAN BIRD DATABASE, WILDLIFE ACT and WILDLIFE CONSERVATION SOCIETY.















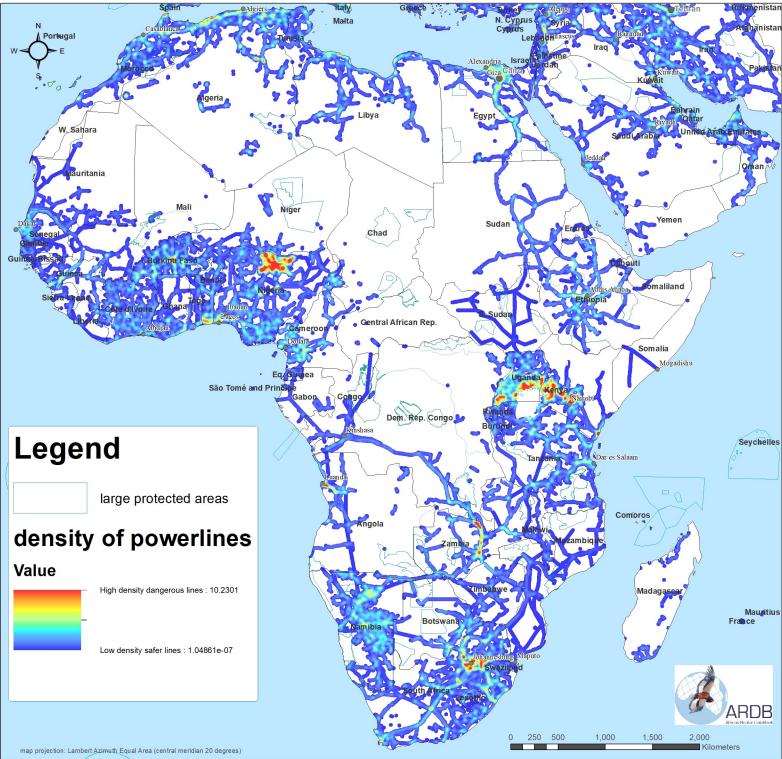






threat map: electrocution



































METHODS:

We accessed new electricity grid data and compared these by intersection to our 2017 sources: Africa infrastructure Country Diagnostic AICD, Foster & Briceno-Garmendia 2010; data on powerlines accessed from OpenStreet Map for North Africa and Arabia (OpenStreetMap contributors 2015); the Digital Atlas of Namibia published by the ACCAIC (And Climate, Adaptation and Cultural Innovation Africa) project. The World Bank data catalogue offers an updated AICD version but we found the Electricity Grid Africa dataset from JRC (Kakoulaki & Moner-Girona 2020) to be the far superior coverage for the continent. These data have been more rigorously obtained from a variety of year work. We used their data in preference but imported our previous data for several locations in Africa and for the Middle East & Eurose where the JRC dataset offers no coverage. As before, all powerlines were given weighting scores. Those without useable attribute information on KV or status were given a median score of 3; existing or under construction powerlines > 40kv carried a score of 2; while powerlines that were planned, missing, proposed, under study and > 40kv were given lowest scores of 1; a maximum score of 5 was attributed to any existing powerlines of voltage < 40kV which are considered to be of greatest threat to replors, and a value of 4 was allocated to <40kV inlies under considered to be of greatest threat to replors, and a value use as the population field and a serior rigid with a page such as a such a

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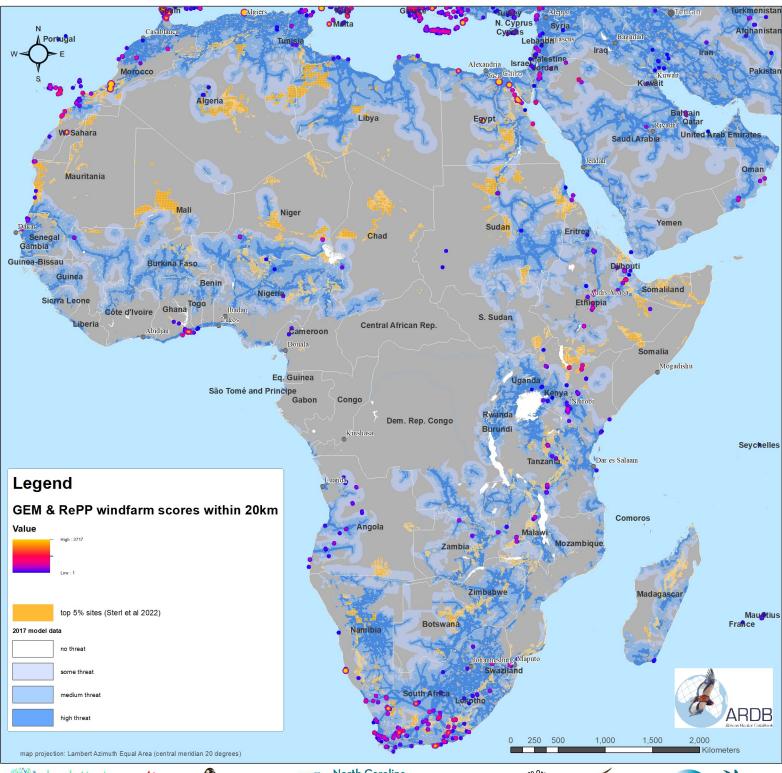






threat map: wind turbine collision risk

































METHODS: In this map we include two new datasets overlaying our original 2017 model layer for turbine collision risk (which is shown in shades of blue). The first is another wind farm suitability model by Sterl et al. (2022) which shows in pale orange the top 5% best areas for wind farm development for each suitable African country. Their model uses more recent and more complete data than our 2017 model and takes account of further aspects of supply and demand e.g. population size of closest city. These areas matched well with our 2017 his suitability predictions. Upon this, in bold localised colours we show empirical data we have compiled on existing, planned and under construction wind farm locations. These locations, along with status and the power capacity of each wind farm are compiled from two sources: Global Wind Power Tracket at 2023 release for global sites) supplemented by RePP data for the African continent (Peters et al. 2023). For each wind farm location we estimated the footprint area of the wind farm on the basis that a 2mw turbine within a wind farm placement usually occupies a footprint of 0.5 kmz. We calculated the radius of the wind farms. We gave the following scores to wind farms on the basis of their status: 5 for existing and operational. Ye runder construction. 3 for planned, proposed; and 1 for retired or not operational. We then converted the polygons to rasters using the score field for value and we summed the scores of all grid cells within a 20km radius to provide an indication of collision risk from these data representing actual wind farms.

wind farms.
Steft. S., Hussain, B., Miketa, A. et al. An all-Africa dataset of energy model "supply regions" for solar photovoltaic and wind power. Sci Data 9, 664 (2022). https://doi.org/10.1038/s41597-022-01786-56 https://www.nature.com/articles/s41597-022-01786-56 data-availability
"Global Wind Power Tracker, Global Energy Monitor, May 2023 release." https://globalenergymonitor.org/projects/global-wind-power-tracker/
Peters, R., Berlekamp, J., Tockner, K. et al. RePP Africa – a georeferenced and curated database on existing and proposed wind, solar, and

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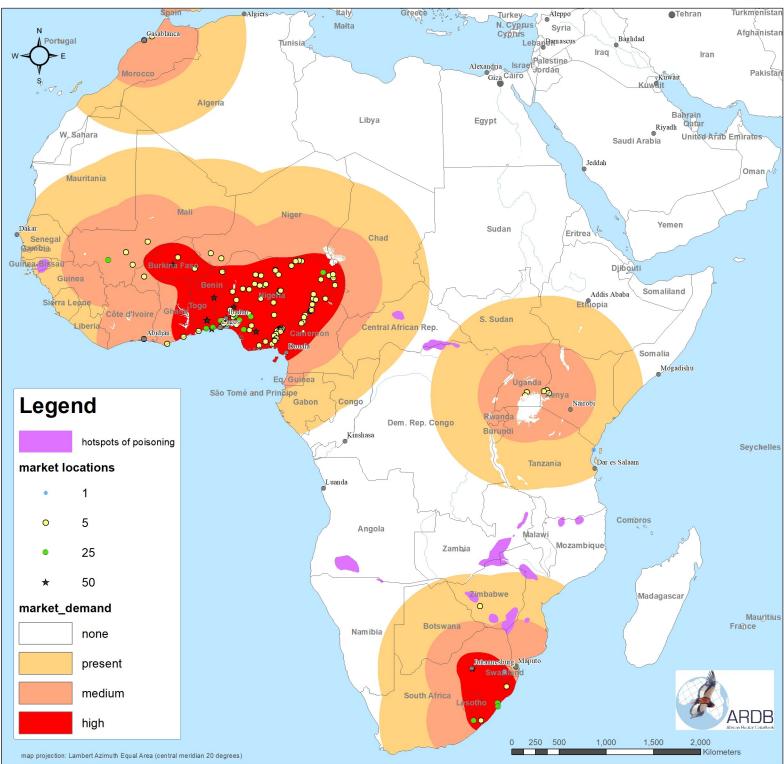






threat map: belief-based use

22 December 2023

































For the original map in 2017 we digitized and georeferenced the locations of 125 traditional medicine markets. Lou Luddington assisted with this work. These were mainly in West & Central Africa provided by Buij et al. (2016) but we supplemented these with several known locations for Southem and Eastern Africa from Williams et al. (2014), McKean et al. (2013) and other sources (Ogada, Thomsett, Monadjem, Pomeroy, Baker & Baker in litt). We also tabulated information on the size of these markets. We do not yet have a systematic way of measuring this but we looked at the number of stalls with vulture partoucts and the number of vultures traded over time periods (McKean et al. 2013). We classed markets as non-trading in vulture parts (weighted 1), small (weighted 5), medium (weighted 25) or large (weighted 50) – these weights were roughly based on the frequency histogram of number of stalls with vulture products. We conducted a kernel density analysis across Africa measuring the density of markets within a 500km radius and using the weights as a population field. The resultant dataset was then reclassified into four threat letvels: 0 = not hreat, 1 = threat present, 2 = medium threat / demand, 7 a high threat / demand, For the 2023 update we also ran the poisoning likelihood model (see method for map T1) but using only records of poisoning where the heads of carcases had been removed and any other indications for belief-based use (n=116). We tried various methods to overcome biases in the recording effort. The latest records (since 1976) from the GRIN database of raptor provided a convent biase layer for incorporation into the maxent model. This run of the model yielded Area Under the Curve (AUC) value of 0.903 for test records (15% of data were held back for evaluation purposes). Of the explanatory variables sheep density tow). There are some indications that poisoning for belief-based used is more prevalent in highly populous areas, close to protected areas. The model results were granular so we ran a focal stati

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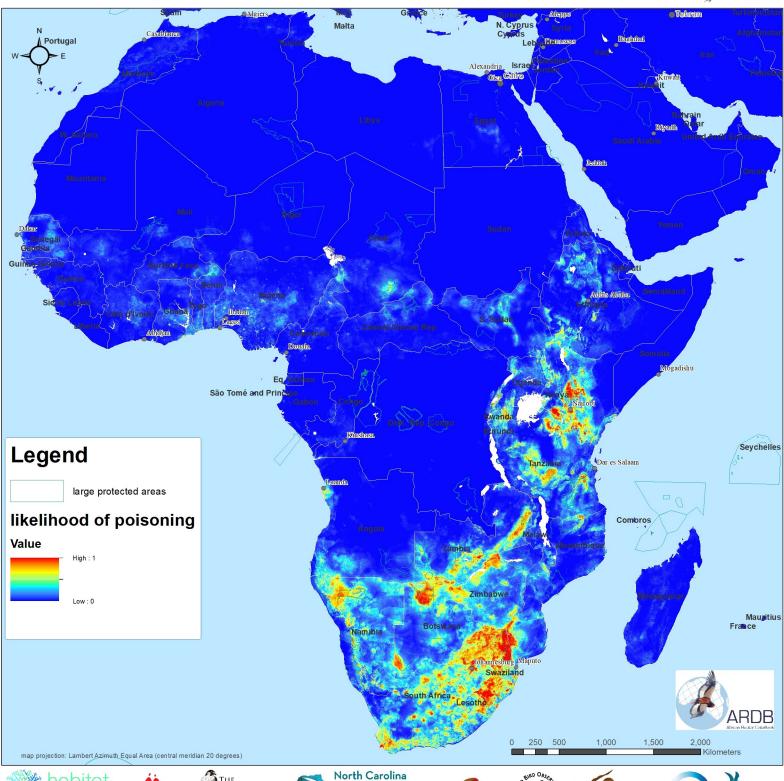






poisoning likelihood

































METHODS: The African poisonings database has been compiled, geo-referenced and maintained by the Endangered Wildlife Trust (Lizanne Roxburgh & Andre Botha), Habitat Info and The Peregrine Fund (Darcy Ogada). Between 2017 and 2023 this database (all species) has grown from 444 to 2369 records. As before we conducted an inductive modelling exercise using the same 97 explanatory variables and Maxent software that were used in the vulture distribution modelling in 2017. The result was a polationing likelihood ray that Indicated very strong likelihood of poisoning in Southern and Eastern Africa with little likelihood in Western Africa where several thousand vultures have been reported poisoned at a single location. We were concerned that there may be observer bias in the poisoning location gathering exercise because many of these data are collected by wildlife rangers and many are trained and deployed in southern and Eastern regions. We addressed the possibility of observer bias in the data in a number of reveal of the presentation layer for observer effort from the total number of recent sightings of birds of prey from the GRIN database representing the activities of good observers across the continent. We rarefilled the poisoning locations data by only using records that were at least 10m apart. Our most satisfactory model results were finally obtained by using only the records of vultures and other raptors (884 records) and by restricting the explanatory variables to 50 which represented anthropogenic effects (Southern and Eastern Africa comprise the Afromontane betwith shared physiognomic and climatic factors so these non-anthropogenic effects (Southern and Eastern Africa comprise the Afromontane between the physiognomic and climatic factors so these non-anthropogenic effects (Southern and Eastern Africa to region and the proposal proposal). Of the explanatory variables is No Rosine (also hade the largest contribution to this model (more poisoning in Christian areas, previously ranked third), followed by distanc

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