Using indicator species to detect high quality habitats in an East African forest biodiversity hotspot

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Abstract
Species demanding specific habitat requirements suffer, particularly under environmental changes. The smallest owl of Africa, the Sokoke Scops Owl (Otus ireneae), occurs exclusively in East African coastal forests. To understand the movement behaviour and habitat demands of O. ireneae, we combined data from radio-tracking and remote sensing to calculate Species Distribution Models across the Arabuko Sokoke forest in southern Kenya. Based on these data, we estimated the local population size and projected the distribution of current suitable habitats. We found that the species occurs only in Cynometra woodland with large old trees and dense vegetation. Based on home range sizes and the distribution of suitable forest habitats, the local population size was estimated at <400 pairs. Ongoing selective logging of hard-wood trees and the production of charcoal are reducing habitat quality of which will reduce the low numbers of O. ireneae, and of other specialist forest species, even further. Due to their close connection with intact Cynometra forest, O. ireneae is an excellent indicator of intact forest remnants. In addition, this species is a suitable flagship for the promotion and conservation of the last remaining coastal forests of East Africa.

Keywords Cynometra forest · Habitat quality · Otus ireneae · Population size · Movement behaviour · Radio-tracking · Roosting trees · Species distribution model · Carrying capacity

Introduction
One of the main drivers of global biodiversity decline is habitat destruction (Maxwell et al. 2016) with many tropical ecosystems being transformed into agricultural land for large-scale cash-crop production and subsistence agriculture (Geist and Lambin 2002). Not only...
are natural habitats, species and local populations being lost (Cincotta et al. 2000; Miles et al. 2016), but the quality of remaining habitat is deteriorating, especially for taxa with specific habitat demands (Carignan and Villard 2002). Even forests with a closed tree canopy are being compromised by selective logging and the exploitation of natural resources, such as the collection of dead wood and the poaching of animals (Gatti et al. 2015).

Various strategies have been proposed to identify and manage endangered ecosystems and species. One strategy is to use indicator species to identify intact remnants of endangered ecosystems, and to evaluate their ecological integrity (Carignan and Villard 2002). Thus, surrogate species which represent a specific habitat type and which are highly sensitive to changes of habitat quality may help to identify the best remaining remnants of an ecosystem (Githiru et al. 2007; Murphy et al. 2011; Banks et al. 2014). The explanatory power of indicator species is high for two reasons: first, they can indicate it is likely other sensitive species will inhabit the same ecosystem, and second, because indicator taxa immediately respond to environmental changes, such as the deterioration of habitat quality (Simberloff 1998).

The forests of tropical Africa suffer the highest rates of deforestation in the world (Hansen et al. 2013; Achard et al. 2014), with 93.0% of natural forests of East Africa having been destroyed (Aleman et al. 2018). The quality of small and patchy forest remnants is also decreasing, primarily because of selective hardwood logging, firewood collection, charcoal production, and various infrastructure projects (Gordon and Ayiemba 2003; Habel et al. 2017). Selective logging, for instance, affects tree species composition and changes habitat structure including canopy height and vegetation density, and reduces the availability of many resources needed by specialist species (Gatti et al. 2015).

The Sokoke Scopes Owl (*Otus ireneae*) occurs exclusively in *Cynometra* woodland, which makes up a fraction of the dry coastal forest in East Africa (Virani 1994; Fanshawe 1995; Cuadros-Casanova et al. 2018). Apart from a few very small isolated populations elsewhere (Evans 1997; Virani 2000; Virani et al. 2010), almost the entire population of *O. ireneae* is restricted to the Arabuko Sokoke forest, located north-east of Mombasa in southern Kenya (Virani 1994; BirdLife International 2018). However, the Arabuko Sokoke forest has been subject to the illegal logging of large *Brachylaena* and *Cynometra* trees, which has changed the forest’s structure (Fanshawe 1995; Muriithi and Kenyon 2002). Due to this habitat destruction and the declining habitat quality, the population of *O. ireneae* has declined considerably during the last two decades (Virani 2000; Virani et al. 2010).

In this study, we radio-tracked ten *O. ireneae* individuals in order to analyse the species movement behaviour, and to estimate home ranges. We then combined the radio-tracking data with satellite images in order to calculate Species Distribution Models (SDMs). Based on the habitat requirements derived from the SDMs, we projected the extent of the potentially suitable *O. ireneae* habitat and calculated the carrying capacity of the forest (Fig. 1).

**Methods**

**Study area**

East African dry coastal forest consists of three forest types: *Cynometra* forest (further subdivided into *Cynometra* thicket and *Cynometra* woodland), *Brachystegia* forest, and Mixed forest (formerly named *Afzelia* forest or *Hymenaea-Manilkara* forest; Muriithi and Kenyon 2002). The Arabuko Sokoke forest in southern Kenya is the largest remaining dry coastal
forest block in East Africa, with a total extent of 41,600 ha (Wass 1995; Bennun et al. 1996). Considered an Important Bird Area (IBA), the Arabuko Sokoke forest is listed as the second most important site for bird conservation in Africa due to the high number of bird species, including many forest endemics (Collar and Stuart 1988; Bennun and Njoroge 1999).

**Radio-tracking**

We performed radio-tracking of *O. ireneae* during the transitions between the Wet and Dry Seasons, in July and August 2016–2017. Ten individual owls (all adults, gender could not be determined) were caught with mist-nets set around their roosting trees. We banded each owl and measured their head, wing, tarsus, and body weight. We then attached a radio transmitter (Pip Ag376; Biotrack Ltd, Wareham, UK, with a weight of 1.5 g) as a backpack with a "wing-loop" harness made of natural rubber to each of the owls. The weight of the radio transmitter was less than 2% of the total body weight (mean 50 ± 1.7 g), well below the commonly used threshold (4–5% of body weight; see Kenward 2001).

We conducted radio-tracking with hand-held receivers (AR8200-MK3, AOR Inc. USA) and four-element Yagi antennas. Two observers simultaneously took bearings of tagged owls every 20 min from 6 pm until 6 am using a compass, recording their own position with a hand-held GPS device (Garmin GPSmap 64x). The two observers maintained positions of at least 300 m distant from each other. Data collection was split into two 6 h periods (6 pm till midnight; midnight till 6 am). During the early morning, the tagged owls were followed in order to locate their roosting trees. The exact position of each owl was calculated by triangulating the bearings of both observers. Positions calculated for the first day after tagging were excluded due the potential for atypical movement behaviour. Positions with distances > 1000 m were excluded from the data set as the transmitters had a maximum range of 1000 m. The roosting trees of each individual were identified, and subsequently geographically determined.

**Home range estimation**

Home range sizes of *O. ireneae* individuals were estimated using Minimum Convex Polygons (MCP) and Autocorrelated Kernel Density Estimation (AKDE) for 95% (transitional area) and 50% levels (core area). The MCP home range sizes enabled us to compare our data with the results of previous studies (e.g. Virani 1994). The AKDE method allowed for the more precise estimate of particularly small home range sizes, compared with conventional Kernel density estimators, as it corrects estimates for temporal autocorrelation (here, 20 min intervals; Noonan et al. 2019). The AKDE home range size were calculated by fitting continuous-time movement models (ctmm), which were visually inspected in order to assess the autocorrelation structure of the movement data and selected using the Akaike Information Criterion, corrected for small sample size (AICc). From the best model, the AKDE utilisation distribution and home range sizes were calculated (Fleming and Calabrese 2017). Incremental area analysis was then conducted in order to determine whether home range size reached an asymptote, thus indicating that the number of locations was sufficient. In this study, home range estimates reached an asymptote after 180 to 200 relocations. The analyses were performed using the “Animove-Triangulation” package in QGIS v. 2.0.1; MCP home range estimators were calculated with the R package “adehabitatHR”
Species distribution model

SDMs were created on the basis of calculated positions (fixes) obtained from radio-tracking and the following environmental variables: tree cover, tree height, Normalized Difference Vegetation Index (NDVI), spectral information from blue, Mid-Infrared (MIR), Near-Infrared (NIR) bands, and vegetation type. These environmental variables were selected following observations of the species’ habitat characteristics and behaviour (see Cuadros-Casanova et al. 2018) and are based on land cover data derived from satellite imagery (sources detailed in Table 1). These were resampled to a resolution of 30 m using a nearest neighbour approach (package “raster”, Cran R, Hijmans 2016). Maximum Entropy modelling (MaxEnt v. 3.4.1) was employed to perform SDM (Phillips et al. 2006) using a Jackknife test in order to measure variable importance in the model development. Model performance was evaluated by examining the area under the receiver operating characteristic curve (AUC), where values of 0.5–0.7 were considered low and thus represented poor model performance, whereas values of 0.7–0.9 were considered moderate, and values > 0.9 represented excellent model performance (Phillips et al. 2006). Based on the assumptions that the home ranges of individuals, i.e. pairs, do not overlap, and that O. ireneae primarily occurs as pairs (i.e. two individuals per home range), the following total pair numbers were estimated by combining the home range estimate per individual (E/I) with the potentially suitable habitat size (SHS). Following a bootstrap approach, 20% of the fixes were randomly selected as test records, while the remaining 80% were used to train the model. This procedure was repeated ten times and averaged over all repetitions. The lowest ten percentile training omission threshold accounting for potential spatial triangulation errors was selected as the presence-absence threshold. All computations were performed in QGIS v. 2.0.1, and MaxEnt v. 3.4.0 (Phillips et al. 2006). Finally, the carrying capacity for O. ireneae was estimated based on the spatial expansion of potentially suitable habitats obtained from SDMs, and the mean home range size of the AKDE calculations.

Results

Ten O.ireneae individuals were radio-tracked for, on average, 23.5 ± 1.0 nights (ranging from 20 to 25 nights). The O.ireneae home range size was calculated by MCP95 at 14.2 ± 1.9 ha, and MCP50 at 2.8 ± 0.3 ha. The transitional area calculated by AKDE95 was 18.9 ± 2.2 ha, and the core area calculated by AKDE50 was 3.5 ± 0.3 ha (Table 2). The home ranges of radio-tracked owls did not overlap, or did so only marginally.

MaxEnt models provided reasonable predictive power with AUC test values of 0.77–0.87. The area of suitable habitat was estimated at 5283 ha. Most of the projected suitable habitat distribution was fragmented Cynometra woodland with largest patches in the north-western and south-eastern parts of the Arabuko Sokoke forest (Fig. 2). Estimates of the relative contributions of the environmental variables showed that tree height, NDVI, and NIR explained the most variation (percentages contributed by all variables are presented in Table 3). The Jackknife test indicated the most pronounced drop in explanatory power when omitting tree height. Computing SDMs based on the single variables NDVI and MIR had
Table 1  Ecological variables used to create Species Distribution Models for *Otus ireneae* in the Arabuko Sokoke Forest, Kenya

| Variable description                      | Platform       | Year  | Resolution | Sources                          |
|------------------------------------------|----------------|-------|------------|----------------------------------|
| Tree cover                               | Landsat 7 + ETM| 2000  | 30 m       | Hansen et al. (2013)             |
| Canopy closure (%) for all vegetation taller than 5 m in height |                |       |            |                                  |
| Tree height                              | ICESat         | 2005  | 1 km       | Simard et al. (2011)             |
| Highest points (m) in the forest canopy  |                |       |            |                                  |
| Vegetation type                          | –              | –     | –          | Kelsey and Langton (1984)        |
| Classification of vegetation cover constructed from floristic studies |                |       |            |                                  |
| Vegetation Indices                       | MODIS          | 2018  | 250 m      | Didan et al. (2015)              |
| Enhanced Vegetation Index EVI            |                |       |            |                                  |
| Normalized Difference Vegetation Index NDVI |             |       |            |                                  |
| Surface reflectance                      | MODIS          | 2018  | 250 m      | Didan et al. (2015)              |
| Near Infra-Red (NIR)                     |                |       |            |                                  |
| Mid Infra-Red (MIR)                      |                |       |            |                                  |
| Blue                                     |                |       |            |                                  |

MODIS scene: MOD13Q1.A2018225.h21v09.006.2018241235803
| Species individual ID | A01 | A03 | A04 | A05 | A06 | A07 | B01 | B02 | B03 | B04 | Mean ± SE |
|-----------------------|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----------|
| Tracking period (days)| 25  | 28  | 28  | 21  | 21  | 22  | 25  | 25  | 20  | 20  | 23.5 ± 1.0|
| Start                 | 23 Jun 2016 | 30 Jun 2016 | 30 Jun 2016 | 19 Jul 2016 | 19 Jul 2016 | 19 Jul 2016 | 19 Jul 2016 | 19 Jul 2017 | 19 Jul 2017 | 17 Aug 2017 | 25 Jul 2017 |
| End                   | 21 Jul 2016 | 28 Jul 2016 | 28 Jul 2016 | 15 Aug 2016 | 15 Aug 2016 | 15 Aug 2016 | 15 Aug 2017 | 19 Aug 2017 | 19 Aug 2017 | 19 Aug 2017 | 17 Aug 2017 |
| Number of relocations | 298 | 300 | 300 | 198 | 221 | 222 | 215 | 238 | 231 | 197 | 242.0 ± 13.1|
| MCP95 (ha)            | 19.2 | 26.2 | 12.8 | 15.1 | 6.2 | 11.4 | 10.2 | 9.0 | 12.1 | 19.5 | 14.2 ± 1.9 |
| MCP50 (ha)            | 3.1  | 1.7  | 2.6  | 4.3  | 1.5  | 2.4  | 3.6  | 2.3  | 3.1  | 3.9  | 2.8 ± 0.3  |
| AKDE95 (ha)           | 25.6 | 31.6 | 17.8 | 19.2 | 7.8  | 14.4 | 14.8 | 13.1 | 19.8 | 24.9 | 18.9 ± 2.2 |
| AKDE50 (ha)           | 4.0  | 3.5  | 3.1  | 3.7  | 1.8  | 3.0  | 4.1  | 2.6  | 4.5  | 5.0  | 3.5 ± 0.3  |
the highest explanatory power, whereas vegetation type had the lowest explanatory power (Table 3). Only a fraction of the total Cynometra forest was indicated as being potentially suitable for O. ireneae, with 7613 ha (32.2%) being suitable based on MCP95, 7361 ha (31.3%) based on AKDE95, and 5433 ha (23.1%) being suitable based on AKDE50.

Fig. 1 Home ranges of ten radio-tracked Otus ireneae individuals in the Arabuko Sokoke Forest, Kenya. Light grey: transitional area (AKDE95); dark grey: core area (AKDE50); black triangles: roosting trees, white dots: mean centre.
Finally, the *O. ireneae* carrying capacity of the entire Arabuko Sokoke forest was estimated. The total estimate was 387 pairs across the Arabuko Sokoke forest based on AKDE95, 19 ha E/I and 7361 ha SHS and 544 pairs based on MCP95, 14 ha E/I and 7613 ha SHS.

**Discussion**

**Home range and population size**

The population of *O. ireneae* in the Arabuko Sokoke forest occurs at much higher densities than the only other population for which density measures exist—the Usambara Mountains where there were 1.5–4.0 pairs/km² over an area of 9700 ha (Evans 1997). Nevertheless, the average home range size we obtained for tracked *O. ireneae* based on AKDE95 (18.9 ± 2.2 ha), which is thought to be particularly accurate (Noonan et al. 2019), continue a long-standing increasing trend: Home range size estimates have gone from 12.5 to 14.3 ha/pair in 1973–1979 (Britton and Zimmerman 1979) and 1984 (Kelsey and Langton 1984) to 14.9 ha/pair in 1994 (Virani 1994), 16.1 ha/pair in 2003 and 17.5 ha/pair in 2008 (Virani et al. 2010). The increase in home range size is reflected in the estimate of the
population size: 1025 pairs in 2000 (Virani 2000), 800 pairs in 2008 (Virani et al. 2010) and 387–544 pairs in our study. While such differences in population densities and sizes could be a result of differences in data collection techniques, data quality or in the algorithms applied to estimate home ranges, they could also be caused by a decline in habitat quality.

Table 3 Summary of SDM results depicted from Species Distribution Models

| Model evaluation | MCP95 | AKDE95 | AKDE50 |
|------------------|-------|--------|--------|
| Training AUC     | 0.75  | 0.77   | 0.87   |
| Test AUC         | 0.75  | 0.76   | 0.86   |
| Variable contribution |   |        |        |
| Canopy Cover     | 8.67  | 6.918  | 6.03   |
| Vegetation Type  | 0.13  | 0.10   | 0.02   |
| EVI              | 14.90 | 17.53  | 10.67  |
| MIR              | 12.31 | 10.98  | 8.42   |
| NDVI             | 9.14  | 9.35   | 16.74  |
| NIR              | 13.31 | 11.80  | 11.35  |
| Blue             | 5.99  | 6.28   | 6.30   |
| Tree Height      | 35.55 | 37.05  | 40.48  |

AUC without…

| Variable contribution | MCP95 | AKDE95 | AKDE50 |
|-----------------------|-------|--------|--------|
| Canopy Cover          | 0.75  | 0.76   | 0.86   |
| Vegetation Type       | 0.75  | 0.76   | 0.86   |
| EVI                   | 0.74  | 0.75   | 0.85   |
| MIR                   | 0.75  | 0.76   | 0.86   |
| NDVI                  | 0.74  | 0.76   | 0.86   |
| NIR                   | 0.75  | 0.76   | 0.86   |
| Blue                  | 0.74  | 0.76   | 0.86   |
| Tree Height           | 0.72  | 0.73   | 0.82   |

AUC with only…

| Variable contribution | MCP95 | AKDE95 | AKDE50 |
|-----------------------|-------|--------|--------|
| Canopy Cover          | 0.55  | 0.56   | 0.59   |
| Vegetation Type       | 0.50  | 0.50   | 0.50   |
| EVI                   | 0.62  | 0.62   | 0.69   |
| MIR                   | 0.57  | 0.57   | 0.62   |
| NDVI                  | 0.71  | 0.72   | 0.78   |
| NIR                   | 0.67  | 0.67   | 0.74   |
| Blue                  | 0.61  | 0.62   | 0.67   |
| Tree Height           | 0.68  | 0.68   | 0.75   |

Threshold

| Threshold            | MCP95 | AKDE95 | AKDE50 |
|----------------------|-------|--------|--------|
| 10% training presence| 0.37  | 0.36   | 0.40   |

Given are details on variable contributions with respective AUC values, for MCP95, AKDE95 and AKDE50.
Habitat quality

Increasing home range sizes may be a response to decreasing habitat quality with reduced resources (e.g. food). Wiktander et al. (2001) showed that home range sizes decrease when resources become scarce (e.g. during the dry season). Such environmental changes may also be due to anthropogenic habitat deterioration, which influences habitat structures, species communities, and the availability of resources. This, subsequently impacts individual behaviour (Villard et al. 1999; Banks et al. 2012, 2017; Ndang’ang’a et al. 2013). Previous studies have shown that birds living in degraded environments need to cover larger areas to find suitable food sources when compared with birds living in high-quality habitats, i.e. in non-fragmented environments (Carey et al. 1990; Hansbauer et al. 2008). This might also be the case in the Arabuko Sokoke forest, where habitat quality is decreasing due to selective hardwood logging, charcoal production, and the removal of dead wood (Glenday 2005; Virani et al. 2010; Matiku et al. 2013).

In addition to anthropogenic activities, changes in the density of mega-herbivores such as elephants can also impact ecosystem structure (Villard et al. 1999; Ndang’ang’a et al. 2013; Banks et al. 2012, 2017), species composition and populations of single taxa (Fanshawe 1995; Clergeau and Burel 1997; Barros and Cintra 2009). Previous studies have shown that elephants increase the accumulation of dead wood and the availability of elephant dung, leading to a rise in insect activity and changes to arthropod community composition which then affects higher trophic levels, such as birds (Otieno et al. 2014). In addition, elephants may destroy large trees, which provide important breeding sites to *O. ireneae*.

Indicator and flagship species

Our SDMs indicated that tree height (availability of large old trees), NDVI (biomass), and vegetation type (*Cynometra* woodland) are the strongest explanatory variables for predicting *O. ireneae* occurrence within the Arabuko Sokoke forest. The SDM demonstrated the restriction of *O. ireneae* to the *Cynometra* woodland forest type, a finding in line with the study by Virani (1994) on *O. ireneae* habitat demands in the Arabuko Sokoke forest. Our data further indicated several roosting places per owl, always located in the centre of the AKDE50 area. Owls return to these roosting places after nocturnal foraging, using different roosting places in an alternating manner. All roosting places were characterised by high *Cynometra* trees and were covered by dense lianas and vines (Camilo Zamora & Ivon Cuadros-Casanova, personal observations). This again underlines the importance of undisturbed *Cynometra* woodland to this sensitive bird species.

Our SDMs reveal that *O. ireneae* is neither homogeneously distributed across Arabuko Sokoke forest, nor homogeneously distributed across *Cynometra* woodland (a fraction of Arabuko Sokoke forest; Fig. 2). The only suitable habitat in the Arabuko Sokoke forest is *Cynometra* woodland with dense vegetation and large, old trees and a high proportion of deadwood. Decreasing habitat quality may reduce the carrying capacity of *O. ireneae*, and may also negatively impact other forest specialist species. With a mean body weight of 50 ± 1.7 g, *O. ireneae* is the smallest owl of East Africa, and its colourful plumage makes it a very charismatic species. Along with being an indicator of intact *Cynometra* woodland forest, makes this owl an excellent flagship species to promote and conserve this endangered biodiversity hotspot.
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