

Estimating conservation metrics from atlas data: the case of southern African endemic birds

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Summary

The robust assessment of conservation status increasingly requires population metrics for species that may be little-researched, with no prospect of immediate improvement, but for which citizen science atlas data may exist. We explore the potential for bird atlas data to generate population metrics of use in red data assessment, using the endemic and near-endemic birds of southern Africa. This region, defined here as South Africa, Lesotho and Swaziland, is home to a large number of endemic bird species and an active atlas project. The Southern African Bird Atlas Projects (SABAP) 1 and 2 are large-scale citizen science data sets, consisting of hundreds of thousands of bird checklists and > 10 million bird occurrence records on a grid across the subcontinent. These data contain detailed information on species' distributions and population change. For conservationists, metrics that guide decisions on the conservation status of a species for red listing can be obtained from SABAP, including range size, range change, population change, and range connectivity (fragmentation). We present a range of conservation metrics for these bird species, focusing on population change metrics together with an associated statistical confidence metric. Population change metrics correlate with change metrics calculated from dynamic occupancy modelling for a set of 191 common species. We identify four species with neither international nor local threatened status, yet for which bird atlas data suggest alarming declines, and two species with threatened status for which our metrics suggest could be reconsidered. A standardised approach to deciding the conservation status of a species is useful so that charismatic or flagship species do not receive disproportionate attention, although ultimately conservation status of any species must always be a consultative process.

Introduction

In a world with biodiversity under increasing threat from anthropogenic activities, it can be difficult to prioritise the limited resources available to conservationists. Big, charismatic species may receive a disproportionate share of resources at the expense of small, nondescript or inaccessible species (Leader-Williams and Dublin 2000). Systematic, objective ways of determining conservation status are thus increasingly important in a world in which conservation is easily driven by emotional rather than logically defensible criteria.

The understanding of bird distributions and how they are changing is crucial for effective bird conservation (Gaston 2003). In order to determine the conservation status of any species, information is needed on range size, population size and population trend (IUCN Standards and Petitions Subcommittee 2014). Furthermore, information on habitat integrity and population fluctuations is also considered where available. However, these data are lacking for most species globally.

We explored ways of facilitating the conservation assessment process in regions of the world with atlas data. Here we focus on southern Africa, defined for this purpose as South Africa, Lesotho and Swaziland. In South Africa alone, there are over 600 species of breeding birds in

a country of 1.2 million km², making biodiversity monitoring a challenging task (Taylor *et al.* 2015). However, the region is increasingly known internationally for its high-quality long-term, large-scale public participation projects. Such citizen science projects make it possible for observations made by many different people to be pooled and analysed as a whole (Cohn 2008). The first and second Southern African Bird Atlas Projects (SABAP₁, 1987–1992, and SABAP₂, 2007–present) are among Africa's biggest biodiversity databases, and provide overviews of avian distribution across southern Africa approximately 20 years apart (Harrison *et al.* 2008). Spatial records in this database show changes in species' distributions (range), and by comparing reporting rates (the proportion of checklists reporting a species - a proxy for relative abundance) between these projects we can estimate population change. This information has been used to examine issues of conservation interest, including the influence on birds of climate change (Walther and Niekerk 2014); identification of non-climatic drivers of range change (Péron and Altwegg 2015a) and changes in timing of migration (Bussière *et al.* 2015), as well as other questions of ecological interest (e.g. Péron and Altwegg 2015b, Péron and Altwegg 2015c, Cooper *et al.* in press). A few publications (e.g. Kemp *et al.* 2001, Lee and Barnard 2015) have used this information to inform conservation decisions for some species, but despite evidence that reporting rate declines are related to observed population decline (Amar *et al.* 2015), there has been little testing of the robustness of these measures and they have not been used formally at a national scale to inform conservation decisions.

We explore the capacity of this atlas project to support conservation status assessment by over-viewing the information that can be obtained from the SABAP projects for 58 southern African endemic and near-endemic bird species. We present population change indices together with a measure of statistical confidence in these, as well as range-size, range-size change and range connectivity metrics. This information should be important for those evaluating the conservation status of these species in southern Africa.

Methods

SABAP data

The SABAP data sets consist of bird lists compiled by birding citizen scientists. SABAP₁ used quarter degree grid cells (15'x15') as the sampling unit, corresponding to standard southern African 1:50,000 topographical maps (Harebottle *et al.* 2007). The SABAP₂ spatial sampling unit is the 5'x5' pentad. There are nine pentads nested within each grid cell, so we aggregated the data of the second phase, SABAP₂, at the quarter degree resolution in order to compare both phases. Birders were asked to submit lists of all species that they saw or heard during visits to grid cells of between two hours and five days. As of 2014, SABAP₂ data existed for over 3,000 grid cells, with country-wide coverage illustrated used in our analysis displayed in Figure 1.

We consider those species with > 70% of range or population within South Africa, Lesotho and Swaziland as near-endemics, as listed by Birdlife South Africa (BSA; Lotz *et al.* 2014). We used this list to obtain the national and global conservation status ('Least Concern' [LC] to 'Endangered' [EN]) for these species. We restrict our analysis to this subset of the > 840 species in the SABAP database, as population trends identified for endemics and near-endemics can be more accurately inferred from this analysis than for species with significant ranges outside the survey area.

Of the 69 endemics on the above list, we consider 58 after excluding those with recent taxonomic splits. Several species in the BSA checklist have been split since SABAP₁ and are represented as two or more species in SABAP₂. We do not consider new southern African species split from species with a combined range that extends beyond the study area. This includes: Hottentot Buttonquail *Turnix hottentottus*, Karoo Thrush *Turdus smithi*, Cape Parrot *Poicephalus robustus*, and the Long-billed Lark *Certhilauda* spp. complex. The data used in this analysis were accessed from the SABAP₂ database over 29–30 May 2014.

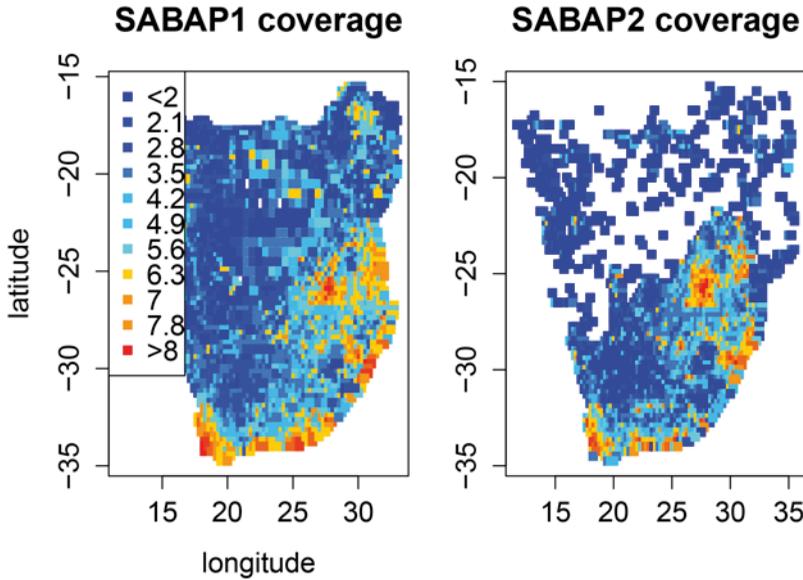


Figure 1. Area and intensity of coverage during the two atlas periods. The colour represents the number of lists reported for each grid cell, with dark grey (red online) high and light grey (blue online) low. SABAP2 coverage is shown to 30 May 2014, as the project is ongoing.

Reporting rate

The reporting rate is the number of times a species was reported in a grid cell divided by the total number of checklists for that grid cell. There is evidence that reporting rates are monotonically related to abundance (Amar *et al.* 2015, Griffioen 2001, Robertson *et al.* 1995). Reporting rate data are publicly available for each species from <http://sabap2.adu.org.za/> and are a useful first step in quantifying changes in abundance (e.g. Huntley *et al.* 2012). For between-atlas period comparisons we select only the subset of data that were sampled on at least two occasions during each atlas period ($n = 2,005$ grid cells). We calculate a summary *reporting rate change* metric based on the average reporting rate across all grid cells for each project for which a species was ever present:

$$(\text{mean SABAP2 reporting rate} - \text{mean SABAP1 reporting rate}) / \text{mean SABAP1 reporting rate}$$

where positive values indicate increase, and negative values indicate decrease. We express this ratio as a percentage. The premise behind this metric is based on the concept of the regression to the mean: while extreme results on a site by site basis certainly exist, the mean across the population should tend towards a stable range of values.

As reporting rate change within a grid cell is undefined if the species was not recorded in that cell during SABAP1, due to division by zero, we create a standardised index of *population change* that allows us to present variation in change across a species range. For each grid cell we calculated the population change metric as follows:

$$\text{SABAP2 reporting rate} / (\text{SABAP1 reporting rate} + \text{SABAP2 reporting rate}) - 0.5$$

This metric returns a value between -0.5 and 0.5, with values > 0 indicating increases, and values < 0 indicating declines (adapted from Amar *et al.* 2010). A population change map for each species

is available as electronic supplementary information, together with reporting rate and range change maps. For an overview of population trends across southern Africa for this set of species, we calculate the mean of population change across all species from within each grid cell as a population change map. We correlate the population change metric against each of the further metrics described below using Pearson's product-moment correlations in R (R Core Team 2015).

As a final visual representation of change, based on list data for the set of endemic birds, we calculate the ratio of lists with a species recorded to lists without that species for each atlas period. This is the presence/absence ratio (presented in Cunningham *et al.* 2016). The log of the mean of these metrics across all grid cells plotted against each other allows a visualization of species that are doing well in SABAP2 compared to SABAP1 as a function of range. We emphasise that species close together on the resulting chart do not necessarily have similar populations, as the reporting rates are influenced by species detectability; for instance, large or vocal species are likely reported more frequently than expected, given density.

The standard statistic for the equality of two proportions (z-score; Underhill and Bradfield 1998) can be used as an index to measure confidence in change in relative abundance that accounts for the number of lists submitted for each grid cell for each period. The following is the formula as described in Underhill and Brooks (2014):

$$Z = \frac{P_2 - P_1}{\sqrt{\left(P(1-P) \left(\frac{1}{n_1} - \frac{1}{n_2} \right) \right)}}$$

where P_1 and P_2 are the reporting rates from SABAP1 and SABAP2 respectively, n_1 and n_2 are the numbers of checklists on which the reporting rates are based, and P , reporting rate, is given by:

$$P = \frac{n_1 P_1 + n_2 P_2}{n_1 + n_2}$$

We calculate the mean of the z-score for the grid cells in a species range as an index of confidence in the direction of population change for each species: large negative values indicate evidence for population decline, large positive values indicate evidence for population increase. Values close to zero indicating unclear status: populations could be declining, increasing or not changing.

Population change metric validation with dynamic occupancy modelling

Treating reporting rate as a proxy for abundance relies on the premise that variation in detection probability is largely due to variation in abundance. This assumption is shared with other abundance estimators that are based on detection / non-detection data (Péron and Altwegg 2015a, Royle and Nichols 2003) and appears to be reasonable for the SABAP data (Huntley *et al.* 2012, Robertson *et al.* 1995). We also assume that the trends in species abundance in areas not well covered (notably the arid central western regions) were similar to those in well-covered areas. As we cannot validate these assumptions and it has been shown that simple metrics can produce biased trend estimates when sampling is not considered (Isaac *et al.* 2014), we test the population change metric described above against 'probability of reporting' change between atlas periods based on 191 common species from dynamic occupancy modelling methods proposed by Bled *et al.* (2013) and presented in Péron and Altwegg (2015a). These models attempt to account for variation in detectability that is a consequence of observer, habitat and season. However, due to the large number of variables these models are unstable for species with low reporting rates and small ranges i.e. most of the species in our set of endemic birds.

We present correlation coefficients at the community level using mean values for each species based on the summary metrics explained above and probability of reporting change for the set of 191 bird species. We also examine correlation between population change scores and z-scores with

probability of reporting change for each of the 20 endemic species at the QDGC level within the set of 191 birds. Lastly, we examine the relationship between the correlation coefficient output for the last analysis with the 20 endemic species with the log-normalised number of QDGCs to examine the influence of range size on these comparisons.

Range and range change

Between atlas periods, ranges of some species expanded while others have contracted. To capture a snapshot of net gain or loss in range, we use the following calculation based on grid cells in which a species has been recorded:

$$(\text{count of grid cells from SABAP2} - \text{count of grid cells from SABAP1}) / \text{count of grid cells from SABAP1}$$

Plotting reporting rate change against range change is useful for gauging how well a species is doing compared to other species.

To exclude that range where perhaps a species was vagrant or possibly incorrectly recorded in SABAP₁, we excluded grid cells that had > 50 lists but only 1 record in SABAP₁ and call this core range. We calculated core range change as above for range change; but this is a stricter measure of range change.

We calculated the total number of grid cells where a species was recorded over both atlas periods. This value multiplied by the approximate area covered by a grid cell, 729 km², we call the species SABAP range, which we consider a surrogate for Extent of Occurrence (EOO; the minimum convex polygon encompassing all known normal occurrences of a particular species). We compare these to ranges from BirdLife International species accounts from the BirdLife Data Zone (<http://www.birdlife.org/datazone/home>) using standard correlation tests in R (R Core Team 2015). We also calculate the area of pentads from which a species has been recorded and treat this finer scale reporting as an indication of Area of Occupancy (AOO; the subset of the EOO where the species actually occurs).

Connectivity index

For each species we calculated a range connectivity score. Each grid cell where a species was recorded was scored for the presence of the species in the four neighbouring grid cells to the north, south, west and east, being those grid cells with greatest surface area contact. The maximum score is four for a grid cell surrounded by other occupied grid cells, while an isolated grid cell will have a score of 0. For each species we record the mean connectivity score across the species range. This index may be influenced by detection probability: a species with a checker-board pattern might occur widely but be hard to detect (e.g. Peregrine Falcon *Falco peregrinus*). As this score is a function of the area of a species range, we correct by dividing the connectivity score by the log of the number of grid cells in which a species occurs.

Results

Population change for southern African endemics in relation to reporting rate and range change

Displaying population metrics for this set of species allows one to examine individual species trends in the broader context of species trends for the region. In the example of what we call southern African endemics, population trends were mixed, with mean reporting rate lower for 33 species, and higher for the remaining 25 (mean population change = -0.06 ± 0.09). There was a correlation between range change and reporting rate change ($t = 6.7$, $P < 0.01$, $df = 56$, Figure 2); as, generally, if a species is no longer reported from a grid cell this is reflected in both metrics.

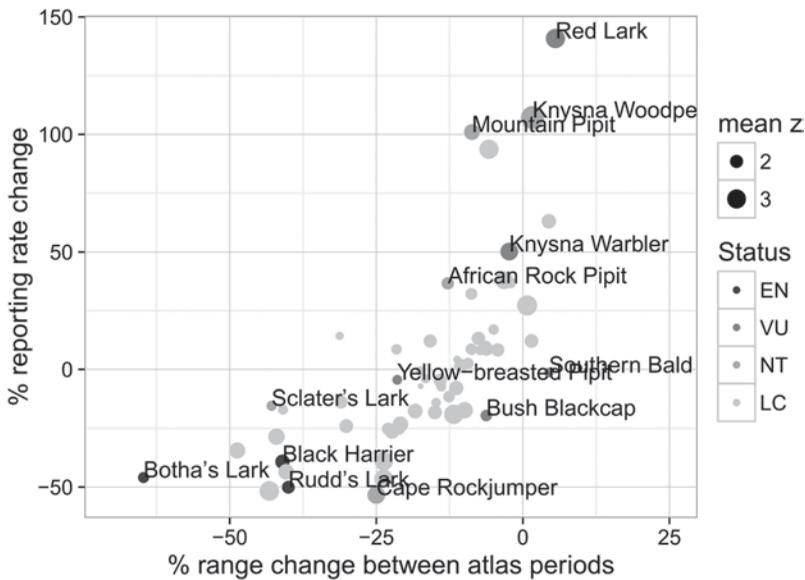


Figure 2. Reporting rate change for 58 South African endemic bird species plotted against change in reported range between SABAP₁ and SABAP₂. Point size represents the absolute value of the mean z-score.

A species with reporting rate change < -30% and range change < -30% (bottom left, Figure 2, Figure 3) may be a species of conservation concern based on IUCN criteria A (population size reduction), where population reduction (measured over the longer of 10 years or three generations) is greater than 30%. Several species with positive range and reporting rate change metrics are still species of conservation concern, as there may be reasons other than population change metrics for considering their status (e.g. population size and fragmentation).

There are four species currently listed as 'Least Concern' that merit further investigation into their conservation status: Ground Woodpecker *Geocolaptes olivaceus*, Drakensberg Rockjumper *Chaetops aurantius*, Sentinel Rock Thrush *Monticola explorator* and Gurney's Sugarbird *Promerops gurneyi* (Figure 3). All these species are associated with upland areas or the grassland biome, as are the three species with existing threatened status in Figure 3. Confidence in population change (mean z-score) of species of conservation concern was lowest for Botha's Lark *Spizocorys fringillaris*, but the total number of grid cells where the species was ever recorded was only 20. The associated range change for Botha's Lark between atlas periods was -64%. The total core range of this species was 15 grid cells, with a range change of -50%: an alarming apparent contraction.

The seven species identified in Figure 3 as species of conservation concern are all species for which the confidence measure (mean z-score) across grid cells was within the lower quartile of values for the total set of species (Table 1). In addition to the above species, those species with large measures of confidence in decrease (negative z-scores) were: Orange-breasted Sunbird *Anthobaphes violacea*, Cape Rockjumper *Chaetops frenatus*, Protea Seed-eater *Crithagra leucop-tera*, Pied Starling *Lamprotornis bicolor* and Grey-winged Francolin *Scleroptila africana*. For these species we are more confident there are population declines possibly as they are associated with areas with large atlas efforts, although the magnitude of these declines may not necessarily meet IUCN criteria for threatened status without concurrent declines in reporting rate. Presence/absence ratios for Protea Seed-eater, Grey-winged Francolin and Pied Starling are low (Figure 4) and appear among the set of species faring most poorly according to this measure.

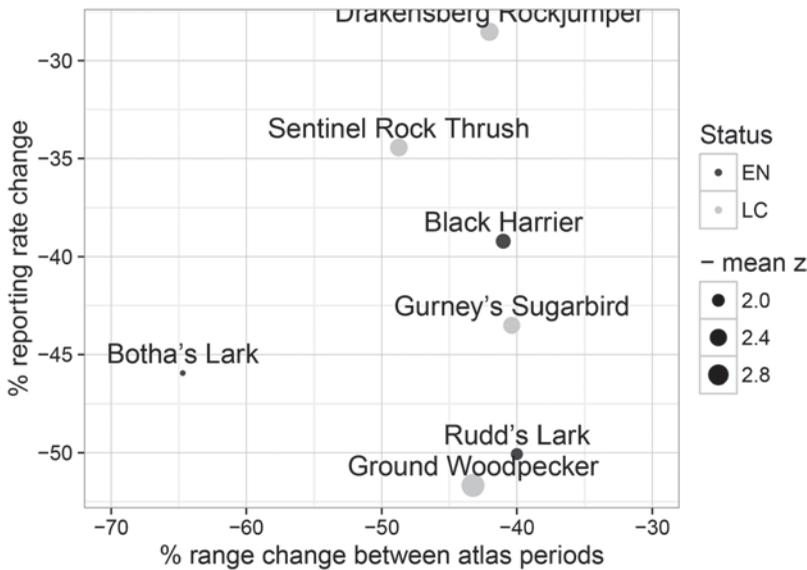


Figure 3. Reporting rate change for seven South African endemic bird species plotted against change in range between SABAP₁ and SABAP₂. This figure shows the lower left hand corner of Figure 2 in more detail - species qualifying as those of conservation concern due to range and population decrease. Size of the points is weighted by mean z-score.

The set of species for which we are most confident of population declines are those associated with grassland and fynbos (a biome restricted mostly to the Western and Eastern Cape provinces (Cowling 1995); Table 1, Figure 5). The overview map of population change suggests endemic species as a whole are faring particularly poorly around Swaziland and north-eastern South Africa. The fynbos biome (a fire-driven ecosystem dominated by shrubs, geophytes and the Restionaceae family; Cowling 1995) in South Africa's most south-western corner is also an area with largely negative trends.

Population change metric validation with dynamic occupancy modelling

For a set of 191 passerine species for which probability of reporting change was calculated from dynamic occupancy modelling (Péron and Altwegg 2015a), at the community level there was a significant positive correlation with both our population change metric ($r_s = 0.46, t = 7.1, df = 189, P < 0.001$) and mean z-scores ($r_s = 0.56, t = 9.3, df = 189, P < 0.001$). However, in the analysis at the species level for the 20 endemic species for which we had occupancy estimates at the grid level, 13 species showed a significant positive correlation between the dynamic occupancy modelling probability of reporting change and population change; while 15 species showed significant correlation with z-scores. Lastly, the size of range seemed to influence this relationship as there was a significant negative correlation between correlation coefficient output from the above analyses and range size for the 20 endemics ($r_s = -0.79, t = -5.3, df = 18, P < 0.001$) suggesting this relationship between occupancy model metrics and our metrics is weak for species with smaller ranges.

Range

The SABAP ranges of endemic and near-endemic species in southern Africa were generally large (> 20,000 km²; an IUCN threshold criteria for determining threatened species status). SABAP range and published EOO values were highly correlated, with those from SABAP lower on average

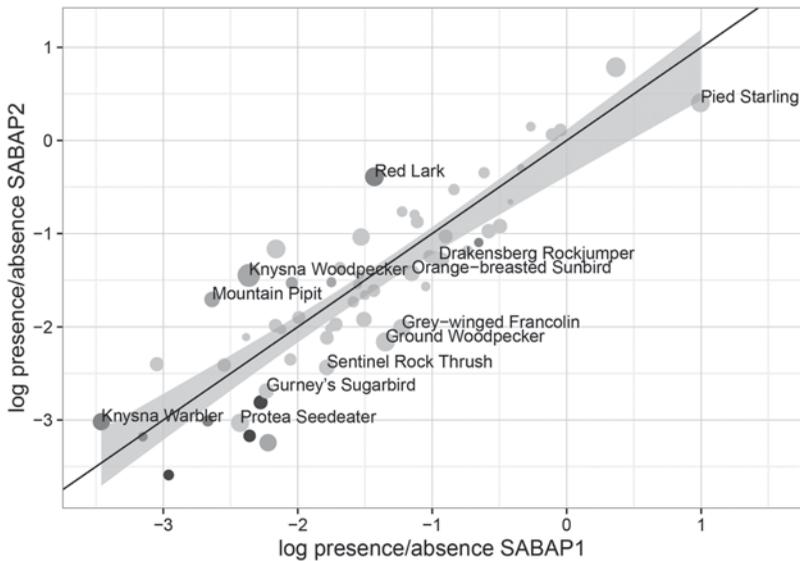


Figure 4. Presence/absence ratios for 58 endemic bird species for each atlas period. Species on the negative end of the x-axis are generally infrequently reported, while those on the positive side are commonly reported: negative values indicate species reported from less than 50% of cells. Shading represents the 95% confidence interval of the regression between the values on the two axes. Species below the 1:1 line (black diagonal) are species reported less frequently in SABAP2. Selected species classified as Least Concern with a lower reporting rate in SABAP2 are labelled, as are selected species with threatened status with higher reporting rates in SABAP2.

(BLI EOO 348,306 km²; SABAP range 291,939 km²; $t = 16.7$, $P < 0.01$, $df = 56$). Only two species had total ranges < 20,000 km²: Botha's Lark and Rudd's Lark *Heteromira fra ruddi*. Overall, occupied area as calculated from pentad data was on average 23.6% of that of SABAP range. Apart from Rudd's and Botha's Larks, only Mountain Pipit *Anthus hoeschi* had the pentad area close to 2,000 km², representing that AOO threshold under which a species might meet conservation status criteria. There was no correlation between population change and SABAP range ($t = 0.1$, $P = 0.91$, $df = 56$); or the pentad area from which a species was recorded ($t = 1.4$, $P = 0.16$, $df = 56$).

Range connectivity

Metrics of connectivity varied widely among the set of endemic bird species. Species identified as those of conservation concern by Lotz *et al.* (2014) dominated the cluster of species with high fragmentation and small ranges, both corrected and uncorrected (Table 1, Figure 6). There was a significant positive correlation between the corrected connectivity score and population change ($t = 2.6$, $P = 0.01$, $df = 56$) with species with negative population change also those species with low connectivity. Pied Starling, a widely distributed arid-zone generalist, had the highest connectivity overall. On the other hand, the arid-zone Cinnamon-breasted Warbler *Euryptila subcinnamomea* had the lowest connectivity score.

Discussion

Citizen science projects like bird atlas projects have an important role to play in biodiversity conservation (Robertson *et al.* 2010). In this article we have shown ways in which species atlas data

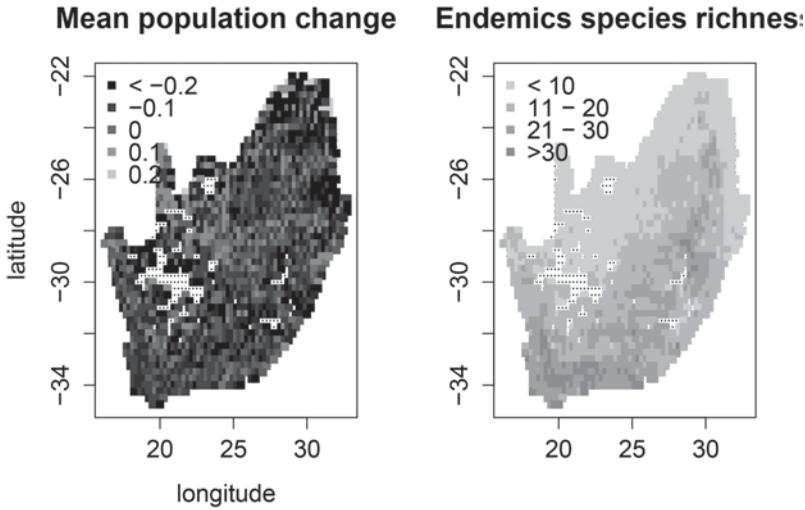


Figure 5. Mean population change across all species within each grid cell (left panel). N for each grid cell is indicated by endemic species richness (right panel). Grids not included in this analysis due to insufficient coverage (<2 lists for both atlas periods) are white with black points.

can be used to develop population parameters that can assist conservation assessment of bird species. However, we regard our analysis as only one approach to be used alongside other lines of evidence when assessing the conservation status of species. In our case study from southern Africa, SABAP2 is a dynamic dataset which facilitates the exploration of numerous ecological and conservation

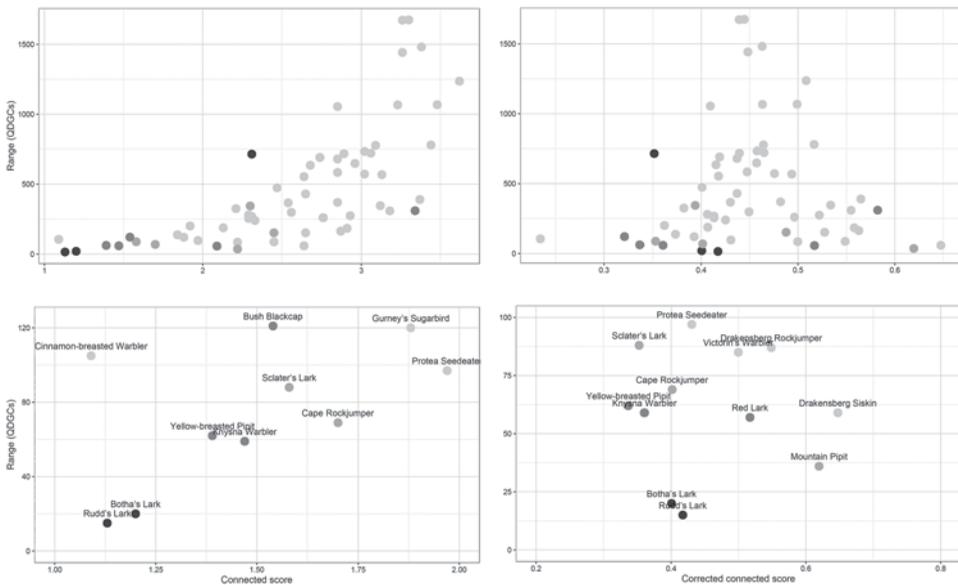


Figure 6. Connectivity (left-hand panels) and corrected connectivity (connected score/log(range); right hand panels) for southern African endemic bird species. The lower two charts are the lower left sections of the upper charts, indicating species with small ranges and low connectivity; QDGC = quarter degree grid cells.

questions. It has been effective as ‘an early warning system’ to alert conservationists of species in trouble (Barnard and Villiers 2012, Lee and Barnard 2012).

Population change

Both Loftie-Eaton (2014) and Péron and Altwegg (2015a) found evidence that the species sets that they examined using atlas data were reported more frequently during SABAP2. The latter study of 191 widely distributed species reported only increases in probability of recording between atlas periods. In contrast to these two studies, several endemic species from our study show evidence for population declines.

In southern Africa, the five worst-faring birds by standardised population change have high affinities to the grassland biome or mountain regions. Of these, Botha’s Lark has been identified as ‘Endangered’. However, we identify four species listed as ‘Least Concern’ that show changes in population status and range size which qualify them as species of conservation concern: Ground Woodpecker, Sentinel Rock Thrush, Drakensberg Rockjumper and Gurney’s Sugarbird. It has been noted that Gurney’s Sugarbird is adversely impacted by inappropriate fire regimes (de Swardt 2010), but there is little published on the other species. It has previously been shown using SABAP data that the species diversity of grassland birds generally, and globally threatened grassland birds in particular, is significantly and negatively correlated with the extent of afforestation (Allan *et al.* 1997). Furthermore, climate envelope modelling suggests that fynbos and grassland bird species are among those most at risk from global climate change (Huntley and Barnard 2012). Our analysis suggests that some species of this biome are in detectable decline over the relatively short period of time between atlas periods.

Several species with grassland affinities show signs of positive population change: Cloud Cisticola *Cisticola textrix*, Melodious Lark *Mirafra cheniana*, Mountain Pipit and Southern Bald Ibis *Geronticus calvus*. However, long term monitoring suggests that Southern Bald Ibis continues to show moderate declines at breeding sites, with a breeding population of < 2,000 pairs (C. Henderson, unpubl. data). This species is a colonial breeder that often forages in groups. Bird atlas data may not be a sufficiently sensitive early warning tool for population declines of charismatic or predictably flocking species, as their abundance is not directly recorded and so even large declines in mean flock size would not be reflected in some cases. However, this does suggest that Pied Starling and Grey-winged Francolin, species fitting this description where declines have been observed, may be worthy of special attention. Mountain Pipit shows an unusual situation that reporting rate change was very positive between SABAP1 and SABAP2, which may be a consequence of small range size, for which these summary metrics become unstable. The range for Mountain Pipit showed moderate decrease, and total current range may be under 20,000 km² within South Africa. Melodious Lark, currently with IUCN red list status ‘Near Threatened’ attributed to moderately rapid population decline, is likely stable. Our analysis supports the most recent local regional ranking of ‘Least Concern’ (Taylor *et al.* 2015).

Species with high affinity for forest generally showed little sign of population decreases. Knysna Warbler *Bradypterus sylvaticus* appears to have expanded its range eastwards despite lower coverage in this part of the species range during SABAP2. This species still exhibits a low degree of range connectivity due to its reliance on isolated forest patches. Knysna Woodpecker *Campethera notata* was the species that fared the best of all the endemics by various criteria. It is classified as ‘Near Threatened’ due to historical loss of range from the east coast, coupled with small estimated populations within protected areas. However, population is currently stable, and in the absence of further threats this species might be classified as ‘Least Concern’.

Two arid zone species showed core range change declines of > 35%: Sclater’s Lark *Spizocorys sclateri* and Black-eared Sparrow-lark *Eremopterix australis*. Cinnamon-breasted Warbler showed reporting rate declines of >30% and very low range connectivity. There are concerns for bird populations of southern Africa’s arid zones, as this area is experiencing dramatic increases in extreme heat events (Cunningham *et al.* 2013). However, the area has been poorly covered during SABAP2 and by contrast Red Lark *Calendulauda burra* showed increases in reporting rate

and moderate range increases. There is thus a continued need for improved atlas coverage of dry regions before concrete conclusions can be made regarding arid-zone species using atlas data.

Birds with affinity for fynbos generally also fared poorly, with Protea Seedeater *Crithagra leucoptera* and Cape Rockjumper showing declining trends. Cape Rockjumper has been identified as vulnerable to warming due to climate change (Lee and Barnard 2015), while Protea Seedeater declines can be attributed to decrease in mature *Protea* sp. and associated food stands as well as nesting sites (Lee and Barnard 2014). Several species with fynbos and grassland affinity appear to be faring poorly, including Black Harrier *Circus maurus*, Ground Woodpecker and Grey-winged Francolin. By contrast, Cape Bulbul *Pycnonotus capensis*, Victorin's Warbler *Cryptillas victorini* and Cape Grassbird *Sphenoeacus afer* show positive population change trends.

Range

The area encompassed by grid cells best fits the IUCN definition of Extent of Occurrence (EOO), defined as that area that can be measured by a minimum convex polygon and which contains all sites of occurrence. However, for species with fragmented range due to poor coverage, this area would currently under-represent the technical definition of EOO. Species with $EOO < 20,000 \text{ km}^2$ may qualify for endangered status if this range is also severely fragmented combined with continuing observed decline in population metrics or extreme population fluctuations. The two species meeting these criteria are Rudd's and Botha's Larks, both currently classified as 'Endangered'.

Range connectivity

Species with small, highly fragmented ranges are traditionally those species most at risk from a conservation perspective (Bolger *et al.* 1991). Sclater's Lark, Cinnamon-breasted Warbler and Black-eared Sparrow-lark are three arid-zone species with low scores. The scores of arid zone specialists may be influenced by poor coverage in the arid western and interior of South Africa. On the other hand, forest species like Chorister Robin-chat *Cossypha dichroa*, Forest Buzzard *Buteo trizonatus* and Knysna Warbler would be expected to have a fragmented distribution as Afromontane forest is a naturally fragmented biome in South Africa. The low connectivity for the upland or grassland species Yellow-breasted Pipit *Anthus chloris* and Sentinel Rock-Thrush is unexpected given the extent of their respective preferred biomes.

Conclusions

We have shown how species atlas datasets, using the example of the Southern African Bird Atlas Project, can be used to extract simple population metrics for use in developing conservation status assessments, even where detailed research on species is unavailable. Our set of southern African endemic and near-endemic bird species shows evidence for population declines among several species. This fits the global pattern that range-restricted species are more vulnerable to patterns of global change. Across southern Africa there is concern that range-restricted species will increasingly have no climate envelope space in which to move (Huntley *et al.* 2012).

While we concentrate on southern African endemic bird species, this region hosts considerable populations of bird species with global conservation status that we have not considered in this assessment, such as Blue Swallow *Hirundo atrocaerulea*, Bearded Vulture *Gypaetus barbatus* and several crane species. While some of the techniques introduced here, such as standardised population change in conjunction with z-scores can be used for these species, there are further caveats to the interpretation of these species, as movements and range changes elsewhere are difficult to account for.

In order to interpret atlas data, greater use should be made of statistics that standardise reporting rates over as wide an area as possible, and greater use should be made of occupancy modelling that accounts for various issues related to detection arising from observer and seasonal effects

(e.g. Bled *et al.* 2013), for species with sufficiently large ranges. The metrics we used in this study rely on the assumption that the probability of detecting a species at a site is dominated by its local abundance and otherwise reasonably constant (Guillera-Arroita *et al.* 2015). In contrast, occupancy- and related models allow for modelling the observation process in more detail (Altwegg *et al.* 2008, MacKenzie *et al.* 2006, Royle and Nichols 2003). However, for species with small ranges compared to the spatial sampling unit, statistically separating the observation process from the biological process can be challenging. In these cases, we argue that comparisons based on raw data can still be useful, provided that they are interpreted with appropriate care. In spite of the difficulties in interpretation of changes in reporting rates between SABAP₁ and SABAP₂, it is likely that if the SABAP₂ results for a species shows decreased reporting rates (or complete absence) over large parts of its range, this reflects genuine range change, as comparisons are more likely to be conservative than to exaggerate increases or decreases (Loftie-Eaton 2014). With SABAP₂ entering its seventh year with consistent reporting for the last five years, this project will in the near future provide information on population change in its own right. There is thus every reason to continue to encourage the citizen scientists who collect these data to continue doing so, and thus add value to one of Africa's largest, and certainly most accessible and vibrant biodiversity databases.

Supplementary Material

To view supplementary material for this article, please visit <https://doi.org/10.1017/S0959270916000307>

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References

- Allan, D. G., Harrison, J. A., Navarro, R., van Wilgen, B. W. and Thompson, M. W. (1997) The impact of commercial afforestation on bird populations in Mpumalanga Province, South Africa—insights from bird-atlas data. *Biol. Conserv.* 79: 173–185.
- Altwegg, R., Wheeler, M. and Erni, B. (2008) Climate and the range dynamics of species with imperfect detection. *Biol. Lett.* 4: 581–584.
- Amar, A., Redpath, S., Sim, I. and Buchanan, G. (2010) Spatial and temporal associations between recovering populations of common raven *Corvus corax* and British upland wader populations. *J. Appl. Ecol.* 47: 253–262.
- Amar, A., Cloete, D. and Whittington, M. (2015) Using independent nest survey data to validate changes in reporting rates of Martial Eagles between the Southern African Bird Atlas Project 1 and 2. *Ostrich* 87: 1–5.
- Barnard, P. and Villiers, M. d. (2012) *Biodiversity early warning systems: South African citizen scientists monitoring change*. Pretoria, South Africa: South African National Biodiversity Institute.
- Bled, F., Nichols, J. D. and Altwegg, R. (2013) Dynamic occupancy models for analyzing species' range dynamics across large geographic scales. *Ecol. Evol.* 3: 4896–4909.
- Bolger, D. T., Alberts, A. C. and Soule, M. E. (1991) Occurrence patterns of bird species in habitat fragments: sampling, extinction, and nested species subsets. *Am. Nat.* 137: 155–166.

- Bussière, E. M. S., Underhill, L. G. and Altwegg, R. (2015) Patterns of bird migration phenology in South Africa suggest northern hemisphere climate as the most consistent driver of change. *Global Change Biol.* 21: 2179–2190.
- Cohn, J. P. (2008) Citizen science: Can volunteers do real research? *BioScience* 58: 192–197.
- Cooper, T. J. G., Wannenburg, A. M. and Cherry, M. I. (in press) Atlas data indicate forest dependent bird species declines in South Africa. *Bird Conserv. Internat.* doi: 10.1017/S095927091600040X.
- Cowling, R. (1995) *Fynbos: South Africa's unique floral kingdom*. Cape Town, South Africa: University of Cape Town.
- Cunningham, S., Madden, C., Barnard, P. and Amar, A. (2016) Electric crows: powerlines, climate change and the emergence of a native invader. *Divers. Distrib.* 22: 17–29.
- Cunningham, S. J., Kruger, A. C., Nxumalo, M. P. and Hockey, P. A. (2013) Identifying biologically meaningful hot-weather events using threshold temperatures that affect life-history. *PLoS ONE* 8: e82492.
- de Swardt, D. H. (2010) Gurney's Sugarbirds in the Lydenburg area. *Environment - People and Conservation in Africa* 2: 42–45.
- Gaston, K. J. (2003) *The structure and dynamics of geographic ranges*. Oxford, UK: Oxford University Press.
- Griffioen, P. (2001) Temporal changes in the distributions of bird species in eastern Australia. PhD thesis. Budoora, Victoria, Australia: La Trobe University.
- Guillera-Aroita, G., Lahoz-Monfort, J. J., Elith, J., Gordon, A., Kujala, H., Lentini, P. E., McCarthy, M. A., Tingley, R. and Wintle, B. A. (2015) Is my species distribution model fit for purpose? Matching data and models to applications. *Global Ecol. Biogeogr.* 24: 276–292.
- Harebottle, D., Smith, N., Underhill, L. and Brooks, M. (2007) *Southern African Bird Atlas Project 2: instruction manual*. Cape Town, South Africa: Animal Demography Unit, University of Cape Town.
- Harrison, J., Underhill, L. and Barnard, P. (2008) The seminal legacy of the Southern African bird atlas project. *South Afr. J. Sci.* 104: 82–84.
- Hockey, P., Dean, W. R. J. and Ryan, P. (2005) *Roberts birds of southern Africa*, 7th edition. Johannesburg, South Africa: Trustees of the John Voelcker Bird Book Fund.
- Huntley, B., Altwegg, R., Barnard, P., Collingham, Y. C. and Hole, D. G. (2012) Modelling relationships between species spatial abundance patterns and climate. *Global Ecol. Biogeogr.* 21: 668–681.
- Huntley, B. and Barnard, P. (2012) Potential impacts of climatic change on southern African birds of fynbos and grassland biodiversity hotspots. *Divers. Distrib.* 18: 769–781.
- Isaac, N. J., Strien, A. J., August, T. A., Zeeuw, M. P. and Roy, D. B. (2014) Statistics for citizen science: extracting signals of change from noisy ecological data. *Methods Ecol. Evol.* 5: 1052–1060.
- IUCN Standards and Petitions Subcommittee (2014) *Guidelines for using the IUCN Red List categories and criteria. Version 11. Prepared by the Standards and Petitions Subcommittee*. Downloadable from <http://www.iucnredlist.org/documents/RedListGuidelines.pdf>.
- Kemp, A., Herholdt, J., Whyte, I. and Harrison, J. (2001) Birds of the two largest national parks in South Africa: a method to generate estimates of population size for all species and assess their conservation ecology. *South Afr. J. Sci.* 97: 393–403.
- Leader-Williams, N. and Dublin, H. T. (2000) Charismatic megafauna as 'flagship species'. *Priorities for the conservation of mammalian diversity: has the panda had its day?* 53–81.
- Lee, A. T. K. and Barnard, P. (2012) Endemic Fynbos avifauna: comparative range declines a cause for concern. *Ornithol. Obs.* 3: 19–28.
- Lee, A. T. K. and Barnard, P. (2014) Aspects of the ecology and morphology of the protea seedeater, *Crithagra leucopterus*, a little-known fynbos endemic. *Afr. Zool.* 49: 295–300.
- Lee, A. T. K. and Barnard, P. (2015) Endemic birds of the Fynbos biome: a conservation assessment and impacts of climate change. *Bird Conserv. Internatn.* 26: 52–68.
- Loftie-Eaton, M. (2014) *Geographic range dynamics of South Africa's bird species*. South Africa: Department of Biological Sciences, University of Cape Town.
- Lotz, C., Allan, D., Bowie, R., Chittenden, H., Cohen, C., Dowsett, B., Gibbon, G., Hardaker, T., Marais, E., Peacock, F., Retief, E., Ryan, P., Smit-Robinson, H. and Taylor, M.

- (2014) *BirdLife South Africa checklist of birds in South Africa*. Available at: <http://www.birdlife.org.za/publications/checklists>, BirdLife South Africa.
- MacKenzie, D. I., Nichols, J., Royle, J., Pollock, K., Bailey, L. and Hines, J. (2006) *Occupancy estimation and modeling*, San Diego, California, USA: Academic Press.
- Péron, G. and Altwegg, R. (2015a) Twenty-five years of change in southern African passerine diversity: nonclimatic factors of change. *Global Change Biol.* 21: 3347–3355.
- Péron, G. and Altwegg, R. (2015b) The abundant centre syndrome and species distributions: insights from closely related species pairs in southern Africa. *Global Ecol. Biogeogr.* 24: 215–225.
- Péron, G. and Altwegg, R. (2015c) Low bird diversity in the Fynbos plant diversity hotspot: Quaternary legacies in the current distributions of passerine birds. *Ecography*: doi: 10.1111/ecog.01176.
- R Core Team (2015) R: A language and environment for statistical computing. Vienna, Austria: R Foundation for Statistical Computing. URL <http://www.R-project.org/>.
- Robertson, A., Simmons, R. E., Jarvis, A. M. and Brown, C. J. (1995) Can bird atlas data be used to estimate population size? A case study using Namibian endemics. *Biol. Conserv.* 71: 87–95.
- Robertson, M. P., Cumming, G. S. and Erasmus, B. F. N. (2010) Getting the most out of atlas data. *Divers. Distrib.* 16: 363–375.
- Royle, J. A. and Nichols, J. D. (2003) Estimating abundance from repeated presence-absence data or point counts. *Ecology* 84: 777–790.
- Taylor, M. R., Peacock, D. S. and Wanless, R. M. (2015) *The Eskom Red Data Book of birds of South Africa, Lesotho and Swaziland*, Johannesburg, South Africa: BirdLife South Africa.
- Underhill, L. and Bradfield, D. (1998) *Introstat*. Cape Town, South Africa: Juta and Company Ltd.
- Underhill, L. G. and Brooks, M. (2014) Preliminary summary of changes in bird distributions between the first and second Southern African bird atlas projects (SABAP1 AND SABAP2). *Ornithol. Obs.* 5: 258–293.
- Walther, B. A. and Niekerk, A. (2014) Effects of climate change on species turnover and body mass frequency distributions of South African bird communities. *Afr. J. Ecol.* 53: 25–35.

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