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Special Article:

Explaining patterns of avian diversity and endemicity:
climate and biomes of southern Africa over the last 140,000 years

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Appendices S1, S2 and S3 – Supporting Information, comprising text, 6 tables and 10 figures.
Abstract

Aim: Test hypotheses that present biodiversity and endemic species richness are related to climatic stability and/or biome persistence.

Location: Africa south of 15°S.

Methods: Seventy eight HadCM3 general circulation model palaeoclimate experiments spanning the last 140,000 years, plus a pre-industrial experiment, were used to calculate measures of climatic variability for 0.5° grid cells. Models were fitted relating distributions of the nine biomes of South Africa, Lesotho and Swaziland to present climate. These models were used to simulate potential past biome distribution and extent for the 78 palaeoclimate experiments, and three measures of biome persistence. Climatic response surfaces were fitted for 690 bird species regularly breeding in the region and used to simulate present species richness for cells of the 0.5° grid. Species richness was evaluated for residents, mobile species (nomadic or partially/altitudinally migrant within the region), and intra-African migrants, and also separately for endemic/near-endemic (hereafter ‘endemic’) species as a whole and those associated with each biome. Our hypotheses were tested by analysing correlations between species richness and climatic variability or biome persistence.

Results: The magnitude of climatic variability showed clear spatial patterns. Marked changes in biome distributions and extents were projected, although limited areas of persistence were projected for some biomes. Overall species richness was not correlated with climatic variability, although richness of mobile species showed a weak negative correlation. Endemic species richness was significantly negatively correlated with climatic variability. Strongest correlations, however, were positive correlations between biome persistence and richness of endemics associated with individual biomes.

Main Conclusions: Low climatic variability, and especially a degree of stability enabling biome persistence, is strongly correlated with species richness of birds endemic to southern Africa. This probably principally reflects reduced extinction risk for these species where the biome to which they are adapted persisted.

Keywords: atmosphere–ocean general circulation model; biome persistence; birds; Cape Floristic Region; climatic stability; Heinrich Events; last glacial–interglacial cycle; species richness.
INTRODUCTION

Global climatic conditions varied substantially during the late Quaternary, extended glacial and short interglacial stages alternating during the last 800–1000 kyr (EPICA community members, 2004). The last 140 kyr spans the final millennia of the penultimate glacial, last interglacial, last glacial and post-glacial (Holocene). Climatic conditions globally were mostly markedly colder than the recent past, although somewhat warmer for part of the last interglacial and to a lesser extent the early Holocene. The last glacial stage also had large millennial climatic fluctuations (Bond et al., 1997; Wolff et al., 2010).

Palaeoecological evidence shows that species responded to these climatic changes mainly by changing their distribution and/or abundance at regional to continental scales (McGlone, 1988; Huntley & Webb, 1989; Graham & Grimm, 1990; Huntley, 1991; Markgraf, 1993; Markgraf & McGlone, 2005; Marchant et al., 2009). Many species’ populations were at times severely reduced as species were extirpated from previously occupied areas, not only in glaciated North America and Europe (Ehlers & Gibbard, 2004), but wherever climatic changes made conditions unsuitable for species’ persistence. These extirpations often markedly reduced the extent and/or spatial continuity of species’ distributions, with consequent population isolation and overall population decreases. In turn, population bottlenecks resulted in loss of genetic diversity (MacPhee et al., 2005; Campos et al., 2010), probably contributing to extinction of some species, whilst increased isolation of sub-populations led to genetic differentiation (Hewitt, 1996, 2001) and probably speciation.

Southern Africa (10 – 35°E, 15 – 40°S) includes the Greater Cape Floristic Region (Allsopp et al., 2014), an area of globally high diversity and endemicity, especially amongst higher plants but also other taxa, including birds (Colville et al., 2014). Even in such unglaciated or minimally glaciated regions (Boelhouwers & Meiklejohn, 2002; Mills et al., 2012), climatic conditions varied substantially over the last glacial–interglacial cycle (Meadows & Baxter, 1999; Partridge et al., 1999; Chase & Meadows, 2007). However, such areas might have high biodiversity because late-Quaternary climatic changes were less than elsewhere, leading to greater persistence of species (Dynesius & Jansson, 2000). Testing this hypothesis requires quantification of the magnitude of late-Quaternary climatic changes and of their relationships with present diversity patterns. Given that species were more likely to be extirpated from more changeable regions, we expect present biodiversity patterns, especially endemic species richness, to reflect patterns in the magnitude of past climatic changes (Huntley et al., 2014). Further, we expect this to be more pronounced for species associated with
particular biomes, especially biomes that experienced marked late-Quaternary changes in location and/or extent.

We aim to test two related hypotheses:

1. That present biodiversity, especially endemic species richness, is greater where the late-Quaternary climate was less variable.

2. That species richness, especially of biome-associated endemics, is greatest where late-Quaternary climate, although variable, allowed persistence of that biome.

**MATERIALS AND METHODS**

**Palaeoclimatic changes**

Results from 78 palaeoclimate experiments and a pre-industrial experiment using a consistent configuration of the Hadley Centre unified model (Singarayer & Valdes, 2010), a fully-coupled atmosphere–ocean general circulation model (AOGCM) including interactive sea-ice (Gordon et al., 2000), were used to derive regional palaeoclimatic estimates. Experiments were performed for time slices from 140 ka to 1 ka at intervals of 1 kyr (1 – 22 ka), 2 kyr (24 – 80 ka and 120 – 140 ka) or 4 kyr (84 – 116 ka); seven ‘hosing’ experiments designed to mimic Heinrich Events H0 – H6 (13, 17, 24, 32, 38, 46 and 60 ka) were also performed. Singarayer and Valdes (2010) provide details of model configuration and derivation of boundary conditions and forcings applied.

For each palaeoclimate experiment, anomalies were computed relative to the pre-industrial experiment for monthly mean temperature, precipitation and cloudiness. Thin-plate splines were fitted to these anomalies at GCM grid cell resolution (2·5° latitude x 3·75° longitude) and used to interpolate them to a 0·5° longitude x latitude grid (using ANUSPLIN, Hutchinson, 1989). Regional palaeoclimate scenarios were obtained by applying interpolated anomalies to observed recent (1961–90) values from the Climatic Research Unit 0·5° dataset (CRU CL 1.0, New et al., 1999). A modified version of the FORTRAN program BIOCLI (Wolfgang Cramer and Rik Leemans) was used to calculate nine bioclimatic variables for each grid cell and palaeoclimate scenario: annual thermal sums above 0°C (GDD0) and 5°C (GDD5); mean temperatures of the coldest (MTCO) and warmest (MTWA) months; an estimate of the annual ratio of actual to potential evapotranspiration (AET/PET); annual total intensity of the wet (TOTWETINT) and dry (TOTDRYINT) season(s); and maximum wet (MAXWETINT) and dry (MAXDRYINT) season intensity (see Huntley et al.,...
2006 for details). Each grid cell was assigned to one of three seasonal rainfall zones for each scenario (Appendix S2) and the frequency with which it was assigned to each zone counted.

Values for each bioclimatic variable were standardized to zero mean and unit standard deviation across all grid cells and time slices; standardized values were used to calculate Euclidean distances for each grid cell between climatic conditions for each time slice, including 1961–90 (hereafter ‘present’), and every other time slice. Three measures of climatic variability of each grid cell were then computed: maximum; mean; and standard deviation of Euclidean distances. Each was calculated for Euclidean distances between present climate and the 78 palaeoclimates, hereafter ‘variability relative to present’, and between the climate of each time slice and that of every other, hereafter ‘overall variability’. Calculations were performed using purpose-written FORTRAN programs. Climatic variability measures were mapped to reveal any spatial patterns (using ArcGIS® and CorelDraw X6®). The frequency with which maximum Euclidean distance corresponded to each time-slice pair was counted across all grid cells.

**Changes in biome extent and location**

Quantitative climatic response surfaces (Huntley et al., 2012) were fitted to model relationships between the extents and locations of nine biomes mapped for South Africa, Lesotho and Swaziland by Mucina and Rutherford (2006, see Fig. S1 in Appendix S2) and present climate. Extent of each biome in each 0·5° grid cell was expressed as a proportion of that grid cell’s land area falling within South Africa, Lesotho and/or Swaziland. Two models were fitted for each biome, each using four bioclimatic variables. All models included MTCO, MTWA and AET/PET, the fourth variable being either TOTWETINT or TOTDRYINT; these variables were selected on the basis of known mechanisms through which they influence vegetation character. Model performance was assessed using the correlation between the proportion of each grid cell’s land area mapped as occupied, and predicted by the model to be occupied under present climatic conditions, by the biome. The model for each biome giving the higher correlation was used to project its potential distribution in southern Africa for each palaeoclimate scenario. Models were fitted and their performance assessed using custom-written FORTRAN programs (based partly on programs written by Patrick J. Bartlein, see Huntley et al., 1989).

Potential total regional extent of each biome for each time slice was computed by summing simulated proportions across grid cells, then normalizing values for each time slice to a sum across biomes equal to
the total number of grid cells (1461). Minimum, maximum and mean potential total extent of each biome were computed as absolute values and as percentages of its simulated present extent.

Potential past biome patterns were mapped by assigning each grid cell to the biome simulated to dominate that cell for that time slice, the dominant biome being that simulated to occupy the maximum extent. The number of grid cells dominated by each biome was also counted for each time slice. Three measures of persistence of each biome were computed for each grid cell across all time slices: (i) the frequency with which the cell was simulated to have been dominated by the biome; (ii) the mean percentage of the cell simulated to have been occupied by the biome; and (iii) the frequency with which the biome was simulated to have been present in the cell, presence being assigned when simulated biome extent in the cell exceeded the threshold value that maximised qualitative goodness-of-fit of its response surface model as assessed using the true skill statistic (Allouche et al., 2006). Frequencies were mapped to reveal any cell(s) where a biome was consistently potentially present or dominant.

Present avian diversity patterns

Bird distribution data were obtained from the Southern African Bird Atlas Project (SABAP, Harrison et al., 1997) at 0·25° resolution for South Africa, Lesotho, Swaziland, Namibia and Zimbabwe, and at 0·5° resolution for Botswana. These data record species’ reporting rates, i.e. proportion of cards returned for a grid cell on which the species was recorded. Quantitative response surface models relating species’ reporting rates to present climate were fitted as described by Huntley et al. (2012), using four bioclimatic variables (MTCO, MTWA, AET/PET and TOTDRYINT) shown previously most often to give the best-fitting model (Huntley et al., 2012).

We fitted models for 697 species regularly breeding in the region, categorised by Hockey et al. (2005) as residents (207 spp.), altitudinal or partial migrants or nomadic (hereafter ‘mobile species’, 402 spp.), or intra-African migrants (88 spp.). Long-distance, mainly inter-continental, migrants were excluded because their richness patterns are unlikely to reflect longer-term stability of regional climate and/or biomes; vagrants were excluded because their occurrences are sporadic. Species endemic or near-endemic to southern Africa (hereafter ‘endemics’) were distinguished from more widespread species, and endemics were categorised according to the biome(s) predominantly used (following Hockey et al., 2005).
Models successfully fitted for 690 species were used to simulate each species’ potential present reporting rate for each 0·5° cell. Rather than reduce these reporting rates to presence–absence and count presences, we summed simulated reporting rates for a cell to obtain a measure of its potential species richness. Sums were calculated for: (i) all species; (ii) all residents (203); (iii) all mobile species (399); (iv) all intra-African migrants (88); (v) all widespread (i.e. non-endemic) species (524); (vi) all endemics (163); two intra-African migrants categorised by Hockey et al. (2005) as near-endemic and one as an endemic breeder, respectively, were excluded; (vii) endemic residents (53); and (viii) endemic mobile species (108). Relationships between contemporary species richness and past climatic variability were assessed using the Pearson product-moment correlation coefficient. Species richness of endemics associated with each biome was also computed for each cell. It was necessary to combine Succulent Karoo and Nama Karoo for this analysis as they are not distinguished by Hockey et al. (2005) when indicating principal habitats used by bird species. In addition, no analysis could be performed for the Indian Ocean Coastal Belt as this is not distinguished by Hockey et al. (2005) as a potential habitat. Relationships between contemporary species richness and biome persistence were again assessed using the Pearson correlation coefficient. In this case correlations were calculated both for all 1461 grid cells in the study area and for 823 grid cells south of 22°S, the northernmost latitude to which data used to fit response surface models for the biomes extended.

RESULTS

Palaeoclimatic changes

Considering maximum climatic differences relative to present (Fig. 1(a)), lowest variability was principally in an area corresponding to the present extent of the Nama Karoo and Fynbos biomes (Fig. S1 in Appendix S1), with outlying relatively stable grid cells along the southern half of the west coast and in a broad area in the east. Greatest variability was principally in the north-west, especially centred upon the area of stabilized dunes spanning the borders of Namibia, Angola and Botswana (Thomas et al., 2000). Overall variability maxima show a similar pattern (Fig. 1(d)), though with a more marked area of low variability extending from south-eastern Namibia to the centre of the Western Cape province of South Africa. Greatest variability was again mainly in the north-west and the northern Kalahari. Considering grid cells simulated to be dominated by each biome under present climate, medians and third quartiles for maximum overall Euclidean distance were smaller for those of the Indian Ocean Coastal Belt, Nama Karoo, Fynbos and Succulent Karoo (Table S3 & Fig. S3 in Appendix S3). Climate over the past 140 kyr is thus simulated to have been more stable in
areas currently occupied by these biomes than in areas occupied by the remaining biomes, although a
minority of grid cells in the Nama Karoo show greater variability. This area of relative climatic stability
broadly corresponds to the year-round rainfall zone both at present and most persistently over the past
140 kyr (Figs. S8 & S9 in Appendix S3).

Across the 1461 grid cells, maximum differences from present climate were for a Heinrich Event stadial in
1288 cases, with maximum difference for almost 23% of grid cells being with respect to H2 (H0 61; H1 221;
H2 331; H3 143; H4 47; H5 259; H6 226). When all pairwise comparisons were examined, 124 ka
contributed to the maximum pairwise difference in 959 cases, the other partner in most cases (942) being a
Heinrich Event, with H2 again the most frequent (H0 14; H1 244; H2 383; H3 48; H4 30; H5 190;
H6 33). Other pairs giving maximum difference for >20 grid squares all included a Heinrich Event; they
were: H0 vs 19 ka (21); H0 vs 22 ka (143); 16 ka vs H5 (23); 17 ka vs H5 (21); 19 ka vs H6 (87);
22 ka vs H5 (48); 28 ka vs H6 (48); and H6 vs 120 ka (21). Overall, a Heinrich Event was one of the pair
giving maximum difference in >98% of cases (1433 of 1461 grid cells).

Mean Euclidean distances with respect to present climate and overall (Figs. 1(b) & (e)) show strikingly
different spatial patterns. They also show, especially for mean differences from present climate, a pattern
distinctly different from that for maximum differences (Fig. 1(a)). Lowest variability with respect to present
(Fig. 1(b)), is focused in the west of Northern Cape province and southern Namibia, the most stable area
being on the coast of south-west Namibia. For overall variability (Fig. 1(e)) the pattern is similar to that for
maximum differences (Figs. 1(a) & (d)), with a broad area of relatively low variability extending from south-
eastern Namibia south to central Western Cape, albeit with relative stability in coastal areas with low mean
differences from present climate (Fig. 1(b)), and greatest variability in north-west Namibia/south-west Angola
and the stabilised dunes spanning the Namibia, Angola and Botswana borders.

The patterns for standard deviation of Euclidean distances (Figs. 1(c) & (f)) again differ somewhat between
variability with respect to present climate and overall variability. Both show areas of lowest variability
centred in the Western Cape and in south-east Namibia. There is also a more limited area of low variability
along the coast of north-east Eastern Cape province and south-west Kwazulu-Natal. For variability with
respect to present climate, a much more extensive area of low variability extends from the east of Eastern
Cape through Kwazulu-Natal to southern Mozambique. In both cases variability is generally lower in South
Africa than in areas to the north, with greatest variability corresponding mainly to areas of greatest variability as assessed by both mean and maximum Euclidean distances.

**Changes in biome extent and location**

Response surface models for biomes all gave high correlations with observed data when used to simulate biome extents in the 0.5° grid cells of South Africa, Lesotho and Swaziland for present climate (Table S1 in Appendix S1). The model using TOTWETINT gave a higher correlation for Desert, Fynbos and Forest, the TOTDRYINT model performing better for the other biomes. The better performing models for individual biomes had correlations of between 0.901 (Forest) and 0.969 (Indian Ocean Coastal Belt). Relatively poor model performance for Forest is because this biome accounts for <0.4% of the land area and does not dominate any grid cells. Biome patterns simulated for the study region as a whole involved spatial extrapolation of the models. However, the simulation for present climate broadly accorded with observed and simulated regional biome patterns (see e.g. Scheiter & Higgins, 2009), no substantially different biomes occurring in the region.

Aside from revealing a general dynamism of biome distributions since 140 ka, maps of dominant biomes (Fig. 2) reveal several systematic patterns. Visually the most striking are the large and complementary changes in extent of Nama Karoo and Grassland, and the simulated extensive presence of Fynbos in the north of the region under many palaeoclimate scenarios. When plotted against age (Fig. 3), the relative extent of most biomes is clearly related to austral summer insolation at 30°S, and also responds markedly to climatic conditions simulated for Heinrich Events. The relative extent of Savanna more closely mirrors global climatic signals such as [CO₂] atm and the δ¹⁸O record from Greenland (Fig. S5 in Appendix S3). Evidence of core areas of persistence of several biomes is seen when the frequency with which each dominates grid cells is mapped (Fig. 4). Accounts of the simulated history of each biome are provided in Appendix S2 and details of the simulations given in Appendix S3 (Tables S4, S5 and S6, Fig. S4).

**Relationships of present avian diversity to climatic variability**

Response surface models fitted for 690 southern African bird species were good, with a high median correlation between observed and simulated reporting rates of 0.585 \( (n = 3964, \ p < 0.001) \). The performance for endemics was better than that for widespread species, with median correlations of 0.630 vs 0.560; this is expected given that virtually all of the geographical, and hence climatic, range of the former
is encompassed by the data used to fit the models. Qualitatively the models also performed well, with median true skill statistic (Allouche et al., 2006) for all (endemic vs widespread) species of 0.842 (0.855 vs 0.835) and median Cohen’s kappa (Cohen, 1960) of 0.720 (0.753 vs 0.710).

Endemic species richness patterns differed from those of species as a whole; species of different movement categories also showed different patterns (Appendix S2; Figs. S6 & S7 in Appendix S3). When considering all species, more than half the correlations between species richness and climatic variability are positive, although mostly of small magnitude (Table 1). Strongly positive correlations indicate greater species richness overall, as well as of residents and intra-African migrants, where mean difference from present climate has been higher over the last 140 kyr. Two strongly negative correlations relate to mobile species, indicating that these species achieve greater richness where standard deviations of climatic differences are lower. Mobile species richness thus tends to accord with expectation, being greater where climatic variability (assessed by standard deviations of Euclidean Distances) has been less.

Correlations for widespread species are all positive, indicating that their species richness tends to be greater where climate has been more variable. In contrast, correlations for endemics not only all have the expected negative sign, but are also on average 3.36 x greater in absolute values than correlations for species as a whole, all achieving high significance (Table 1). For endemics as a whole, and for endemic mobile species, highest correlations are with mean climatic differences from present, whereas for endemic resident species highest correlation is with standard deviation of overall climatic differences. Current species richness patterns of endemics thus accord with expectation, with higher species richness where climate has been less variable over the last glacial–interglacial cycle.

**Relationships of present avian diversity to biome persistence**

The alternative measures of biome persistence showed broadly similar spatial patterns (Fig. 4, Fig. S4 in Appendix S3). All biomes except Forest were persistently present across all 79 time slices in at least a small number of grid cells, whereas only four biomes (Succulent Karoo, Fynbos, Grassland and Savanna) persistently dominated one or more grid cells. Correlations could not be calculated for the first measure of biome persistence for Forest associated species because this biome never dominated any grid cells. Simulated patterns of species richness for endemics associated with each biome showed the expected general spatial patterns (Fig. S7 in Appendix S3). However, because many species are not restricted to a
single biome, areas with lower numbers of biome-associated species are much more extensive than the area of occurrence of that biome.

For six of the biomes that could be analysed, all correlations were, as expected, strongly positive (Table 2). Strongest correlations were for Grassland, Karoo and Desert, with lower values for Forest, Fynbos and Albany Thicket. Highest correlations for Karoo and Desert were with their frequency of occurrence across the study area as a whole, whilst for Forest the highest correlation was with its frequency of occurrence south of 22°S. For Grassland, Fynbos and Albany Thicket highest correlations were with their mean percentage occupied south of 22°S, Fynbos, as expected, showing the strongest increase in correlations when analysis was constrained to the southern part of the study area because this excludes those areas in the north often simulated as occupied by Fynbos (Fig. 2), but remote from the area simulated as consistently dominated by this biome (Fig. 4(d)) and also consistently disjunct from the latter area.

Thus for these six biomes the evidence strongly supports our second hypothesis, that present species richness of biome-associated endemics is highly correlated with degree of persistence of the biome in that grid cell during the last glacial–interglacial cycle.

Savanna was the exception to this pattern. For the study area as a whole correlations were significant and negative; correlations for the area south of 22°S, although positive, were weak. Simulated persistence of Savanna was greatest in the north and east (Fig. 4(g), Fig. S4(g) in Appendix S3), whereas the area richest in savanna-associated regional endemics lies mostly south of 22°S and centrally in terms of longitude (Fig. S7(g) in Appendix S3), overlapping only to a limited extent with areas of high persistence of the biome.

**DISCUSSION**

Previous studies have advanced a wide range of hypotheses to account for global and/or regional patterns of biodiversity and/or of endemism (see e.g. Dynesius & Jansson, 2000; Gaston, 2000). These hypotheses can broadly be categorised into those that explain patterns in terms of present environment versus historical factors. Present patterns of productivity, climate and topographic diversity all correlate with biodiversity patterns in at least some regions (Rahbek & Graves, 2001; Jetz & Rahbek, 2002; Thuiller et al., 2006), but present environment attributes alone cannot generally account for present biodiversity patterns (Jetz et al., 2004). Historical factors, in contrast, can often account for these patterns more completely (Huntley, 1993; Araújo et al., 2008; Voelker et al., 2010; Sandel et al., 2011; Huntley et al., 2014). Historical factors may
operate over a range of time scales, from millions of years (e.g. Linder & Hardy, 2004; Voelker et al., 2010) through the multi-millennial glacial–interglacial cycles of the Quaternary (e.g. Huntley, 1993; Dynesius & Jansson, 2000) to the millennial climatic fluctuations of the last glacial stage (Huntley et al., 2014).

We focus upon the most recent glacial–interglacial cycle, including millennial fluctuations characteristic of glacial stages, and test two alternative hypotheses to account for present patterns of avian species richness in southern Africa: (i) present biodiversity, especially endemic species richness, is greater where late-Quaternary climate was less variable; and (ii) species richness, especially of biome-associated endemics, is greatest where late-Quaternary climate, although variable, allowed persistence of that biome.

Unfortunately the generally sparse independent evidence of past climates and biomes in the region, prior to the Holocene (Chase & Meadows, 2007), severely limits assessment of the extent to which simulated palaeoclimate and biome patterns accord with observations. There is neither a sufficiently dense spatial coverage of such evidence, nor sufficient independently-dated or stratigraphically continuous evidence for the last glacial stage and last interglacial, to support systematic assessment of our modelling results. Furthermore, as Huntley et al. (2014) discussed, at least some of the palaeoclimatic evidence is contradictory. Nonetheless, the AOGCM results are in overall agreement with those few records that are available, with some striking examples of detailed agreement (Huntley et al., 2014; Singarayer & Burrough, 2015).

Comparison of the palaeoclimate simulation results with data from marine sediment cores collected in the adjacent oceans (e.g. Kim et al., 2003) supports this conclusion. Similarly, when expressed in terms of seasonal rainfall patterns, there is broad agreement between the palaeoclimate scenarios (Appendix S2, Figs. S8–10 in Appendix S3) and palaeoenvironmental evidence (Chase & Meadows, 2007).

Although we modelled only the influence of past climate on biome distribution, well-documented glacial–interglacial changes in atmospheric carbon dioxide concentration ([CO$_2$]$_{atm}$) (e.g. Petit et al., 1999) would have shifted the competitive balances between woody and herbaceous plants (Bond & Midgley, 2012), and between C3 and C4 species (Polley et al., 1993). This probably most directly affected the distribution of the savanna biome that is defined by the relative dominance of grasses versus trees. Our models probably over-predict the spatial extent of savanna and under-predict that of grassland when [CO$_2$]$_{atm}$ was lower during glacial stages. We explore here only the role of climate, however, because sufficiently credible dynamic vegetation models incorporating plant growth forms and disturbance regimes needed to investigate the interacting effects of changes in climate and [CO$_2$]$_{atm}$ are not yet available. Whilst progress is being made,
more work is required both on representing tree–grass interactions and resulting biome structure under varying $[\text{CO}_2]_{\text{atm}}$, and especially on developing representations of shrub and succulent growth forms that dominate some southern African biomes.

Difficulties in validating our simulations of palaeoclimates and limitations of our approach to modelling past biome distributions notwithstanding, their reliability is supported by our results (Tables 1 & 2). No consistent pattern emerged with respect to relationships between overall species richness (widespread plus endemic species) and climatic stability. Fewer than half of the correlations had the expected sign, and only two with the expected sign achieved high significance (Table 1). However, the patterns for endemics were clear and consistent with both our hypotheses. Species richness of endemics was generally higher in areas with less variable climate over the past 140,000 years (Table 1). We did not expect weaker correlations for endemic resident species than for all endemics or for mobile endemics. However, a relatively higher proportion of resident than mobile endemics occurs in the Savanna biome than in other biomes (compare Fig. S6(e) & (f) in Appendix S3). The correlation of Savanna-associated endemic richness with biome persistence is also weaker than for other biomes (Table 2). Savanna, however, is likely to have responded more than other biomes to the lowered $[\text{CO}_2]_{\text{atm}}$ during the glacial, with resulting greater equatorward contraction meaning it was likely to have been extensive mainly north of our study region at the last glacial maximum (Elenga et al., 2000). This greater sensitivity and equatorward contraction would offer scope for, and favour adaptation of, regional Savanna-associated endemics to track biome shifts. Similarly, past range shifts and/or northward contractions likely account also for the lack of the expected relationship in the case of widespread species. Indeed evidence of higher present richness in more climatically variable areas is consistent with such a response. Among mobile species as a whole the relationship between richness and climatic stability is generally weaker. This may indicate that such partial and altitudinal migrants, and nomads, persisted by making local movements, paralleling their responses to seasonal and inter-annual climatic variability and exploiting the opportunities offered by topographically diverse landscapes. Such contrasting responses to climatic change pose challenges for biodiversity conservation.

There were generally stronger correlations between biome persistence and richness of biome-associated endemics (Table 2). This indicates that it is not primarily the absolute magnitude of climatic variability that leads to reduced current richness of endemics, but the extent to which climatic conditions have changed sufficiently to result in biome replacement. Persistence of a biome in a given geographical location is likely
to reduce extinction risk amongst biome-associated endemics, whereas biome replacement will frequently cause local extinction. If extensive and repeated over time, this will likely reduce the overall number of endemics associated with the biome. Furthermore, a combination of biome persistence and moderate climatic variability may favour the evolution of new biome-associated endemics (Midgley et al., 2005). Whilst our results for birds will not necessarily apply also to other taxonomic groups in southern Africa, evidence that biome-associated species in other groups may also have persisted where biomes persisted, elsewhere showing range shifts as climatic conditions changed (Tolley et al., 2014), suggests that our results may be more general.

Our results increase concerns about the potential for future biodiversity losses resulting from biome replacements. Given projections that more than half Earth’s land area may experience climatic changes by 2100 sufficient to cause biome replacement (Williams et al., 2007), we conclude that future biome replacement resulting from anthropogenic climatic change is likely to result in substantial biodiversity losses worldwide. Avoiding such losses will require a combination of vigorous mitigation measures, so as to limit the magnitude of climatic change as far as possible, and active conservation measures designed to sustain species’ populations and facilitate their spatial responses to changing conditions.
ACKNOWLEDGEMENTS

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REFERENCES


Supporting Information

Additional Supporting Information may be found in the online version of this article:

**Appendix S1:** Supplementary tables and figures

**Appendix S2:** Supplementary text

**Appendix S3:** Supplementary tables and figures of results

Biosketch

Brian Huntley is a palaeoecologist, ecologist and biogeographer with research interests in the interactions between species, ecosystems and their changing environment. His work encompasses studies of the palaeoecology and palaeoenvironments of the Quaternary, of present ecosystems and biogeographic patterns, and of the potential impacts of anthropogenic global change on species and ecosystems. His research has considered a range of taxonomic groups, from plants to extinct Quaternary mammals, and various ecosystems, from Arctic tundra to fynbos. He has a particular interests in birds and climatic change, and in the development of conservation strategies informed by research into how species and ecosystems respond to environmental changes.

BH and GFM conceived the study. PJV and JSS performed the palaeoclimate simulations. BH carried out the assessment of climatic variability, biome and species’ modelling, analysis of results, preparation of figures and drafting of the manuscript. All authors commented upon and contributed to the final version of the manuscript.

Editor: Jon Sadler
Table 1: Species richness and climatic variability

Table 2: Endemic species richness and biome stability
Table 1: Species richness and climatic variability

<table>
<thead>
<tr>
<th>Species group</th>
<th>Overall climatic difference</th>
<th></th>
<th>Climatic difference with respect to present climate</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Maximum</td>
<td>Mean</td>
<td>S.D.</td>
</tr>
<tr>
<td>All species</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Overall</td>
<td>-0.006</td>
<td>0.017</td>
<td>-0.070</td>
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<tr>
<td>Residents</td>
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<td>0.110</td>
<td>0.001</td>
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<tr>
<td>Mobile species</td>
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<td>-0.051</td>
<td>-0.116</td>
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<tr>
<td>Intra-African migrants</td>
<td>0.043</td>
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<td>-0.024</td>
</tr>
<tr>
<td>Widespread species</td>
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<td>0.089</td>
<td>-0.015</td>
</tr>
<tr>
<td>Endemic species</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Overall</td>
<td>-0.263</td>
<td>-0.344</td>
<td>-0.247</td>
</tr>
<tr>
<td>Residents</td>
<td>-0.143</td>
<td>-0.202</td>
<td>-0.164</td>
</tr>
<tr>
<td>Mobile species</td>
<td>-0.279</td>
<td>-0.358</td>
<td>-0.252</td>
</tr>
</tbody>
</table>

Pearson correlation coefficients between present simulated avian species richness and six measures of climatic variability over the past 140 kyr for 0.5° grid cells in southern Africa (n = 1461, bold entries indicate values for which p < 0.0005). Climatic variability is expressed as the maximum, mean or standard deviation of standardised Euclidean distances between climatic conditions calculated either for all time slices, including the present, relative to one another (overall climatic difference), or for all palaeoclimatic time slices relative to the present. High variability of climate thus corresponds to high Euclidean distance values, whereas a high degree of stability of climate is indicated by low Euclidean Distance values. Correlation coefficients are shown for all species, for all species categorised according to seasonal movements, for widespread (i.e. non-endemic) species, for all endemics and for endemics categorised according to seasonal movements (note that Intra-African migrants cannot by definition be endemic or near-endemic to southern Africa). Negative correlations indicate high species richness associated with low climatic variability (i.e. low Euclidean distance values, and hence high climatic stability) and vice versa.
Table 2:  Endemic species richness and biome stability

<table>
<thead>
<tr>
<th>Biome</th>
<th>All of southern Africa</th>
<th>South of 22°S</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Frequency of dominance</td>
<td>Mean percentage occupied</td>
</tr>
<tr>
<td>Desert</td>
<td>0.474</td>
<td>0.673</td>
</tr>
<tr>
<td>Karoo</td>
<td>0.743</td>
<td>0.751</td>
</tr>
<tr>
<td>Fynbos</td>
<td>0.326</td>
<td>0.180</td>
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<tr>
<td>Albany Thicket</td>
<td>0.318</td>
<td>0.471</td>
</tr>
<tr>
<td>Grassland</td>
<td>0.722</td>
<td>0.762</td>
</tr>
<tr>
<td>Savanna</td>
<td>-0.247</td>
<td>-0.367</td>
</tr>
<tr>
<td>Forest</td>
<td>–</td>
<td>0.299</td>
</tr>
</tbody>
</table>

Pearson correlation coefficients between present simulated avian species richness of endemics using each of seven regional biomes in southern Africa as their principal habitat and three measures of the simulated stability of those biomes over the past 140 kyr for 0.5° grid cells (entries in italics are values for which \( p > 0.025 \); \( p < 0.0005 \) for all other values). Correlations are presented both for the entire study region \( (n = 1461) \) and for the area south of 22°S that corresponds approximately to the region from which the data used to construct models relating biome extents to climate were available \( (n = 823) \). The three measures of biome stability are: the frequency, across all time slices, with which the biome dominates the grid cell (i.e. is simulated to extend over a greater percentage of the land area of the grid cell than does any other biome); the mean extent of the biome in the grid cell, across all time slices, expressed as the percentage of the area of land in the grid cell simulated to be occupied by the biome; and the frequency, across all time slices, with which the biome is simulated to be present in the grid cell. Positive correlations indicate high species richness associated with high stability of the associated biome.
FIGURES

Figure 1: Climatic variability in southern Africa since 140 ka
Maximum (a & d), mean (b & e) and standard deviation (c & f) of Euclidean Distances between present climate and 78 palaeoclimate time slices (a–c), and of all possible pairwise Euclidean Distances (d–f) between the 79 climates, for each 0.5° grid cell. Deep orange indicates low climatic variability (i.e. minimum Euclidean Distances) whereas deep blue indicates high climatic variability (i.e. maximum Euclidean Distances). Mapping of Euclidean Distance values to the colour scale in each case was designed to provide the clearest representation of the spatial pattern of climatic variability.

Figure 2: Simulated biome distributions for southern Africa
Maps show the biome simulated to dominate each 0.5° grid cell for a set of time slices selected to show the extreme contrasts corresponding to time-slice pairs most frequently giving maximum climatic differences, as well as to illustrate responses to both orbital and sub-orbital forcing. The 22, 46, 70, 92, 116 and 138 ka time slices correspond approximately to austral summer (December–January) insolation maxima at 30°S, whilst 11, 32, 60, 82, 104 and 124 ka correspond approximately to austral summer insolation minima at 30°S. The H0, H1, H2, H5 and H6 simulations of Heinrich Events, at close to 13, 17, 24, 46 and 60 ka respectively, reflect maximal millennial climatic contrasts relative to conditions simulated using only orbital and other ‘slow’ forcing factors.

Figure 3: Relative extent of each biome in southern Africa over the past 140 kyr
Simulated extent of each biome for each palaeoclimate scenario, relative to its extent for 1961–90, plotted against age. Also shown are the austral summer (December–January) insolation at 30°S, computed following Laskar et al. (2004), and atmospheric CO₂ concentration, derived as a composite from the Law Dome (Etheridge et al., 1996, 0–0.940 ka), Taylor Dome (Indermühle et al., 1999, 1.020–11.103 ka) and Vostok (Barnola et al., 1987, 12.930–140.430 ka) Antarctic Ice Cores. Grey lines indicate Heinrich Events 0–6.

Figure 4: Frequency with which biomes were simulated to dominate grid cells
Shading indicates for how many of the 79 time slices examined, including the present, climatic conditions in each grid cell result in simulated dominance of a given biome. (a) Desert; (b) Succulent Karoo; (c) Nama Karoo; (d) Fynbos; (e) Albany Thicket; (f) Grassland; (g) Savanna; and (h) Indian Ocean Coastal Belt. (Note: Forest was not simulated as the dominant biome in any grid cell for any time slice.)
Maximum (a & d), mean (b & e) and standard deviation (c & f) of Euclidean Distances between present climate and 78 palaeoclimate time slices (a–c), and of all possible pairwise Euclidean Distances (d–f) between the 79 climates, for each 0.5° grid cell. Deep orange indicates low climatic variability (i.e. minimum Euclidean Distances) whereas deep blue indicates high climatic variability (i.e. maximum Euclidean Distances). Mapping of Euclidean Distance values to the colour scale in each case was designed to provide the clearest representation of the spatial pattern of climatic variability.
Maps show the biome simulated to dominate each 0.5° grid cell for a set of time slices selected to show the extreme contrasts corresponding to time-slice pairs most frequently giving maximum climatic differences, as well as to illustrate responses to both orbital and sub-orbital forcing. The 22, 46, 70, 92, 116 and 138 ka time slices correspond approximately to austral summer (December–January) insolation maxima at 30°S, whilst 11, 32, 60, 82, 104 and 124 ka correspond approximately to austral summer insolation minima at 30°S. The H0, H1, H2, H5 and H6 simulations of Heinrich Events, at close to 13, 17, 24, 46 and 60 ka respectively, reflect maximal millennial climatic contrasts relative to conditions simulated using only orbital and other ‘slow’ forcing factors.
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