



Zucchetto, M. F., N. S. Da Silveira, V. M. Prasnewski, T. Arnhold, T. Sobral-Souza, and J. B. de Pinho. 2021. Effects of future climate change on the geographical distribution of a bird endemic to South American floodplains. Avian Conservation and Ecology 16(2):29. <https://doi.org/10.5751/ACE-01893-160229>

Copyright © 2021 by the author(s). Published here under license by the Resilience Alliance.

Research Paper

Effects of future climate change on the geographical distribution of a bird endemic to South American floodplains

Mayara F. Zucchetto¹ , Natalia S. Da Silveira², Victor M. Prasnewski¹, Tatiane Arnhold¹, Thadeu Sobral-Souza³ and João B. de Pinho¹

¹Postgraduate Program in Ecology and Biodiversity Conservation, Department of Botany and Ecology, Institute of Biosciences, Federal University of Mato Grosso (UFMT) - Cuiabá, ²Department of Biodiversity, Institute of Biosciences, São Paulo State University (UNESP), ³Macroecology and Biodiversity Conservation Lab, Department of Botany and Ecology, Institute of Bioscience, Federal University of Mato Grosso (UFMT) - Cuiabá

ABSTRACT. Future climate change poses an immediate threat to biodiversity. Species are shifting their geographic distribution in direct response to climate change. Species around the globe are expected to redistribute poleward and/or to high elevations. However, in the Neotropical region, the effects of climate change on species distribution are still unclear. Here, we aimed to infer the current and future geographical distribution of an insectivorous bird species endemic to South American floodplains, the Mato Grosso Antbird (*Cercomacra melanaria*). We also estimated shifts in the variables of elevation, longitude, latitude, and slope based on forecasting. Current and future distribution in different climate scenarios were predicted using ecological niche modeling (ENM). A non-parametric Kruskall-Wallis test followed by Dunn's post hoc test were carried out to determine if dependent variables (elevation, slope, latitude, and longitude) differed between climate scenarios. Our findings suggest that future climate change will have negative effects (range contraction) on the distribution of the Mato Grosso Antbird. The decrease in the species' geographical range will cause it to shift to higher elevations, latitudes, longitudes, and slopes. The species should tend to disperse toward more suitable areas located in highlands close to wetlands, such as the headwaters of the Upper Paraguay River Basin (UPB), and toward the Andean highlands. However, this species inhabits strictly forested environments covering a small home range and its mobility capacity is limited. It is possible that the species will not be able to achieve future geographical redistribution, thus increasing its risk of extinction.

Effets des futurs changements climatiques sur la répartition géographique d'un oiseau endémique des plaines inondables d'Amérique du Sud

RÉSUMÉ. Les futurs changements climatiques constituent une menace immédiate pour la biodiversité. Les espèces modifient leur répartition géographique en réaction directe aux changements climatiques. On s'attend à ce que les espèces du monde entier se redistribuent vers les pôles ou les hautes altitudes. Cependant, dans la région Néotropicale, les effets des changements climatiques sur la répartition des espèces ne sont toujours pas clairs. Dans le présent article, nous avons tenté de déduire la répartition géographique actuelle et future d'une espèce d'oiseau insectivore endémique des plaines inondables d'Amérique du Sud, le Grisin du Mato Grosso (*Cercomacra melanaria*). Nous avons également calculé les changements sur le plan des variables d'altitude, de longitude, de latitude et de pente en fonction des prévisions. La répartition actuelle et future selon différents scénarios climatiques a été prédite en utilisant la modélisation de niche écologique (MNÉ). Nous avons réalisé un test non paramétrique de Kruskall-Wallis, puis un test post hoc de Dunn, pour déterminer si les variables dépendantes (altitude, pente, latitude et longitude) différaient entre les scénarios climatiques. Nos résultats indiquent que les futurs changements climatiques auront des effets négatifs (contraction de l'aire) sur la répartition du Grisin du Mato Grosso. La contraction de l'aire de l'espèce se traduira par des déplacements vers des altitudes, des latitudes, des longitudes et des pentes plus élevées. Nous nous attendons à ce que l'espèce ait tendance à se disperser vers des zones plus propices situées dans les hautes terres près de milieux humides, comme les sources du bassin supérieur du fleuve Paraguay (BSP), et vers les hautes terres andines. Toutefois, cette espèce fréquente des environnements strictement forestiers, a un petit domaine vital et une capacité limitée de se déplacer. Il est donc possible que l'espèce ne soit pas en mesure de se redistribuer géographiquement dans le futur, augmentant du même coup son risque de disparition.

Key Words: *Cercomacra melanaria*; *Mato Grosso Antbird*; *niche modeling*; *Pantanal*; *Thamnophilidae*; *wetland*

INTRODUCTION

Species are becoming extinct at alarming rates around the world (Dirzo and Raven 2003), with more severe losses occurring in the tropics (Crutzen 2006, Dirzo et al. 2014, Ceballos et al. 2015). Future climate change and natural habitat fragmentation have been identified as the main factors causing the biodiversity crisis (Pecl et al. 2017, Araújo et al. 2019). There is a consensus regarding the negative effects of climate change on biodiversity. Species tend to shift their geographical distribution in response to climate change (Berg et al. 2010). Globally, species are dispersing toward the poles (Pecl et al. 2017) in search of suitable habitats (Parmesan and Yohe 2003, Colwell et al. 2008). The impacts of the geographical redistribution of species have reached ecosystems and their functions, leading to the formation of new biological communities with species compositions that have never previously co-occurred (Pecl et al. 2017). However, the effects of climate change on species distribution in the tropics remain poorly understood.

In response to future climate change, species can be redistributed following latitudinal and altitudinal patterns in search of climatically suitable regions. Latitudinal redistribution tends to occur mainly toward the poles in response to changes in temperature and precipitation rates (Parmesan and Yohe 2003, Root et al. 2003, Chen et al. 2009, Sunday et al. 2012), whereas altitudinal redistribution toward highlands occurs because of rising temperatures and reduction of glacier ice volume in mountain regions (Francou et al. 2003, Seimon et al. 2007). However, geographic and topographic redistribution may occur concurrently (Díaz-Almeyda et al. 2017). Highland species will suffer more with these effects because they will need to disperse to higher latitudes and/or altitudes, which will become geographically scanty on the planet (Şekercioğlu et al. 2008). Likewise, tropical species will tend to modify their distribution to highlands and/or higher latitudes (Kelly and Goulden 2008, Fei et al. 2017, Pecl et al. 2017). The latter is highly concerning when it comes to threatened species, which are typically characterized as having the smallest distribution ranges.

Altitudinal redistribution may result in two secondary patterns of distribution pertaining to slope and longitude (Santillán et al. 2020). Slope redistribution involves shifting species distribution from plains (less rugged terrain) to mountains (more rugged terrain). In regions with extreme topography (e.g., the Andean region), species redistribution may also follow a longitudinal pattern, but toward highlands (Jankowski et al. 2013, Santillán et al. 2020). In addition, mountains may act as a dispersion barrier, thereby particularly affecting species with limited dispersal ability (Graham et al. 2010). Thus, species dispersal will tend to occur in areas with gentle slopes. Thus, to understand the effect of climate change on species distribution, it is essential to test this effect on latitudinal, longitudinal, and slope redistribution simultaneously.

Some groups and/or species considered good biological indicators of climate change can be useful tools to understand the effects of species redistribution and how it is expected to occur in climate change scenarios (Wormworth and Şekercioğlu 2011, Şekercioğlu et al. 2012). To this end, such species must be known and their occurrence must be well documented (Peterson et al. 2011). Thus, estimates of distribution and tolerance to habitat characteristics

should be as realistic as possible, enabling the identification of priority areas and more effective conservation strategies.

In this regard, it is significant that birds currently represent the most well-known and understood group, even in climate change research (Wormworth and Şekercioğlu 2011, Şekercioğlu et al. 2012). Bird species distributions are strongly linked to seasonal changes in temperature and precipitation, making them some of the best indicators of climate change (Şekercioğlu et al. 2012, Pearce-Higgins and Green 2014). In addition, this group has the advantage of having millions of birdwatchers engaged in citizen science everywhere, contributing to the compilation of extensive data sets (Kinzelbach 1995, Lee and Nel 2020, <https://www.ebird.org>, <https://www.wikiaves.com.br>).

According to future climate simulations, bird species tend to spend more energy on thermoregulation as temperature and seasonality increase, thus compromising survival and decreasing fitness (Walsberg 1993, Crick 2003). Indirectly, climate warming may give rise to gradual changes in the availability of food resources and suitable habitats (Şekercioğlu et al. 2012). In this context, bird species that are more sensitive to future climate change may become extinct and/or shift to apparently more suitable regions that may not have specific habitats for their occurrence and/or for the maintenance of their population (Devictor et al. 2008).

The Neotropical region is home to a high diversity of bird species, of which 1919 occur solely in Brazil (Piacentini et al. 2015). Floodplains such as the Pantanal are host to most of this diversity (582 species; Nunes 2011), providing refuge for resting, feeding, and nesting of migratory and resident birds. However, there is a knowledge gap about how climate change will affect Neotropical wetlands and the species that live in them. Therefore, we selected for this study one of the few bird species endemic to the Pantanal, the world's largest seasonal floodplain. *Cercomacra melanaria* (Ménétriès 1835) belong to the family Thamnophilidae, popularly known as the Mato Grosso Antbird (Fig. 1). The species is resident, strictly insectivorous, its behavior is inconspicuous, and it inhabits the dense understory of humid tropical forests along waterways (Zimmer et al. 2020). The Mato Grosso Antbird's dispersal ability is limited by its shape, which prevents it from flying long distances and, hence, from crossing non-forested areas (Pinho et al. 2006, Nóbrega et al. 2017, Sheard et al. 2020). The species' known distribution is restricted to floodplains and wetlands in the Pantanal region of Brazil and Bolivia (Zimmer et al. 2020). The endemism of the species to the Pantanal floodplain means that its conservation status may be under threat from future climate change. Therefore, forecasting changes in the distribution patterns of restricted-range species in future scenarios is relevant to conservation efforts. The loss of a single species can have severe consequences that may result in the loss of ecosystem services and functions essential for entire ecosystems (Cardinale et al. 2012).

Our purpose was to determine the future potential distribution of the Mato Grosso Antbird, estimating its geographical shift in latitude, longitude, elevation, and slope. We hypothesized that the species' future potential distribution would shift toward higher altitudes and latitudes, as well as longitudinally westward into regions with higher slopes (Parmesan and Yohe 2003, Şekercioğlu et al. 2008, 2012, Pecl et al. 2017; Fig. 2). At higher longitudes and latitudes, the elevations will be higher, for example the Andes

Fig. 1. Mato Grosso Antbird (*Cercomacra melanaria*) occurrence records in the Pantanal. These occurrence records were used in ecological niche modeling.

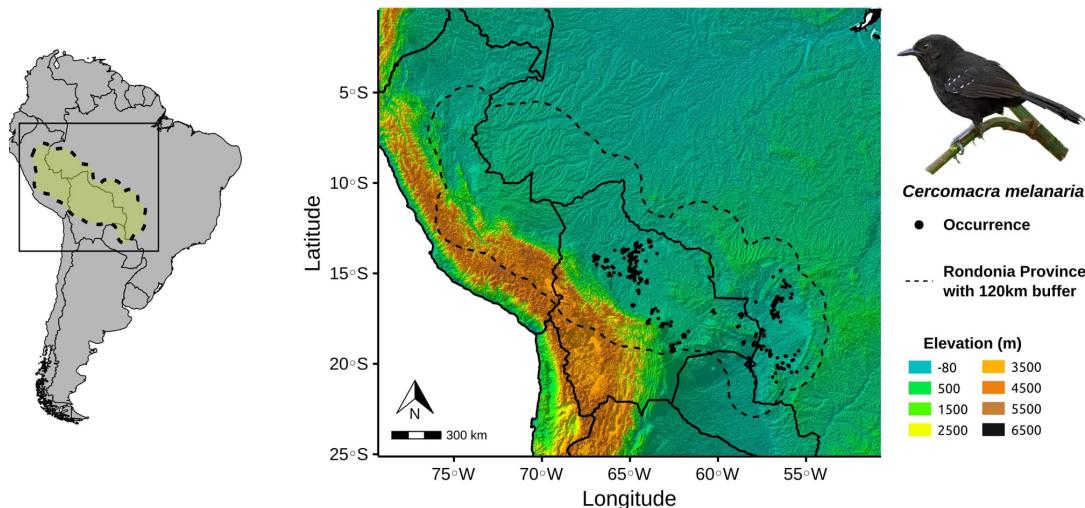
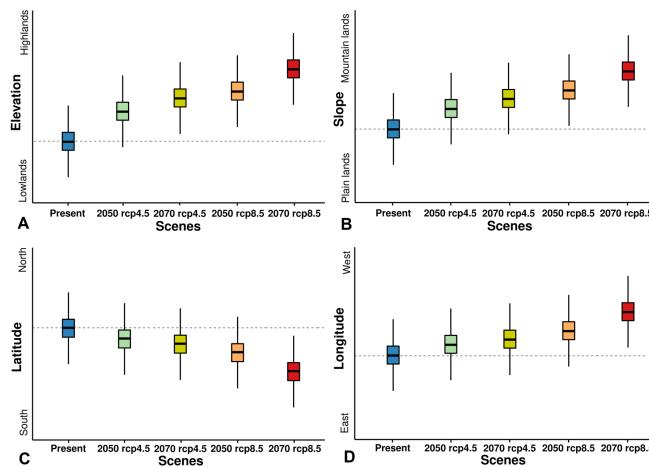


Fig. 2. Predicted habitat suitability of the variables elevation, slope, latitude, and longitude in different scenarios (present, year 2050, year 2070) and different concentrations of greenhouse gases (Representative Concentration Pathway; RCP 4.5 and RCP 8.5) for the Mato Grosso Antbird (*Cercomacra melanaria*). The dashed line indicates the median of the present scenario.



(Jankowski et al. 2013, Santilln et al. 2020), and environmental conditions should be more climatically suitable for the species in the future.

METHODS

Study area

The Mato Grosso Antbird inhabits the Brazilian and Bolivian Pantanal floodplains located in the Upper Paraguay River Basin (UPB; Fig. 1), specifically, in semi-deciduous forest areas and in

the Humid Chaco in Bolivia and in the extreme north of Paraguay. Dispersion is an essential factor in the potential distribution of the species. Hence, this factor must be taken into account in the construction of distribution models, the delimitation of study areas, and the area of calibration of the Mato Grosso Antbird model. In this study, we considered the species' low dispersal ability and its preference for humid forests in low-lying floodplains, i.e., < 200 m (Pinho et al. 2006, Bernardon et al. 2014, Sheard et al. 2020, Zimmer et al. 2020). We assumed a delimited area covering the range of dispersion of existing populations (BirdLife International 2018, Zimmer et al. 2020), reflecting the biogeographic barriers and abiotic tolerance of the species.

The study area was defined based on the biogeographical province of Rondônia (Morrone 2014), covering all known occurrences of the species in southwestern and central Brazil, southern Peru, northwestern Bolivia, and northern Paraguay. Given that the Mato Grosso Antbird also occurs in transition zones of biogeographic provinces and considering the diffuse boundaries that delimit this biogeographical province (Morrone 2014), a buffer zone of approximately 120 km was added to the boundaries of the Rondônia province (Fig. 1). This background area encompasses all known species' occurrences and potential areas for historical dispersion of species, following two criteria for background selection when we applied ecological niche modeling (ENM) approaches (Barve et al. 2011).

Occurrence data

We surveyed known occurrence data using two distinct methodologies: (1) information from citizen science databases and digital databases: eBird (Sullivan et al. 2014), Global Biodiversity Information Facility (GBIF 2019), and iNaturalist (iNaturalist 2019) and (2) occurrence data provided by the Museo Nacional de História Natural, Colección Boliviana de Ornitológica, Universidad Mayor de San Andrés (UMSA) and directly by researchers in the region (Dennis Camacho at the Asociación Boliviana de Ornitológica, Miguel Angel Montenegro at the Museo de Historia Natural Noel Kempff Mercado, and

Isabel Gomez and Sofia Lothesberger at the Universidad Mayor de San Andres). Occurrence records without date and site information, duplicate data, and centroid coordinates of municipalities were excluded. We compiled a total of 302 occurrence records for the 1970–2019 period, which were mapped onto a 2.5-minute grid (approximately 4.5 x 4.5 km resolution at the equator; Appendix 1).

Environmental variables

To characterize background environmental space, we used all the 19 bioclimatic variables available in the WorldClim v. 1.4 database (Hijmans et al. 2005), with cell resolution size of 20.25 km². Because these variables are derived from temperature and precipitation data, they are intercorrelated, requiring a variable selection process to decrease multicollinearity (Peterson et al. 2011). Here, we applied a factorial analysis (Revelle 2020, see Sobral-Souza et al. 2015) with maximum rotation that resulted in four selected variables: Temperature Seasonality (Bio 4), Mean Temperature of Wettest Quarter (Bio 8), Precipitation of Wettest Quarter (Bio 16), and Precipitation of Driest Quarter (Bio 17; Appendix 2). The factor analysis involved all the background points, enabling the selection of the environmental variables based on the space that was modeled.

It is important to highlight that Worldclim 1.4 uses the 1950–2000 time series to encompass temporal variations in climatic conditions. The climatic variation of the last 20 years (2000–2020) is not yet quantified in available climatic masks. Future climate predictions inferred using Intergovernmental Panel on Climate Change (IPCC) data are up to 2100 and project those last 20 years through interpolation. That is why we decided to keep the species occurrence data until 2019.

To test the effects of climate change on potential species distribution, we used future climate scenarios for 2050 (the mid-point for the 2041–2060 period) and 2070 (the mid-point for the 2061–2080 period), with two contrasting emission scenarios (IPCC 2014): (1) IPCC Fifth Assessment Report-Coupled Model Intercomparison Project, Phase 5 (AR-CMIP5) / Representative Concentration Pathway (RCP) 4.5 (lower energy demand) and (2) IPCC AR-CMIP5 / RCP 8.5 (i.e., maximum power requirement and balanced emissions from fossil fuels and non-fossil fuels). We also used three different Atmosphere-Ocean General Circulation Models (AOGCMs) in the future climate models simulations: CCSM4, MRI, and MIROC (<https://esgf-node.llnl.gov/projects/cmip5/>).

Niche-based models building

Ecological niche modeling (ENM) approaches were used to predict current and future potential distributions of the Mato Grosso Antbird. Niche-based models calculate relationships between known occurrence data and environmental predictors (usually climate) to represent tolerable conditions for species survival and to infer habitat suitability values in areas where species occurrence is still unknown (Franklin 2010, Peterson et al. 2011).

To build ecological niche models, we assumed niche stability over time (Peterson et al. 2011, Stigall 2012). It is difficult to predict how species will respond to climates that do not currently exist because their response depends on their adaptive capacity and evolutionary rate (Peterson et al. 2002, Prieto-Torres et al. 2020),

which are not yet known. Therefore, to predict the species' potential distribution in current and future climate scenarios, we used an ensemble forecasting method based on the use of different mathematical algorithms and future climate simulations to model niche species. The combined use of different algorithms usually increases the reliability of models because it considers a wide range of distribution patterns (Araújo and New 2007, Diniz-Filho et al. 2009), and the ensemble approach is currently the one most recommended for niche-based modeling (Araújo et al. 2019, Hao et al. 2019). Hence, we used four algorithms: (1) Bioclim (Envelope Score; Nix 1986), (2) Domain (Gower Distance; Carpenter et al. 1993), (3) Maximum Entropy (MAXENT v3.3.3k; Phillips and Dudík 2008), and (4) Support Vector Machines (SVM; Tax and Duin 2004). The entire niche-based analysis was performed using the R packages "dismo" and "kernlab" (Karatzoglou et al. 2004, Hijmans et al. 2015).

To evaluate the models, we divided the occurrence points into two subsets by means of bootstrapping: 75% of occurrence points were allocated to training and 25% to testing. Because training and testing points are subsets of the same point-occurrence data, we randomized the two subsets 10 times to minimize spatial structure in training and testing datasets, thereby enabling less biased evaluations. Thus, we built 10 replications of each algorithm, totaling 40 models (10 replications X 4 algorithms). The models were built using the current climate scenario and projected for each future scenario.

The model outputs were then transformed into binary maps based on threshold values, which were calculated using maximum sensitivity and specificity. These thresholds values maximize the accuracy of presence-absence and have proved to be effective in presence-only modeling (Liu et al. 2016). To evaluate the models, we calculated the values of true skill statistics (TSS). Because TSS values are threshold dependent, we estimated the values of maximum sensitivity and specificity (Liu et al. 2016). TSS values range from -1 to 1 (Allouche et al. 2006). Values that are negative or close to zero indicate that the forecasts are no different from randomly generated models, whereas forecasts with values close to 1 indicate good models, and values above 0.5 are adopted to indicate the most suitable models (Allouche et al. 2006).

The final habitat suitability maps were built using the frequency ensemble approach. To this end, we overlapped the binary maps pertaining to each algorithm (replicates) using their respective threshold values, and then added up the maps of the same algorithm and between the algorithms to obtain the final consensus map showing the frequencies of occurrence predicted by all the models. The values in the cells indicate the frequencies of occurrence in all the generated models, ranging from 0 to 40 (4 algorithms x 10 repetitions). To calculate final species suitability maps for each temporal and emission scenario, we divided the values of frequencies of occurrence by the total number of generated models (4 algorithms x 10 repetitions) resulting in maps with values from 0 (low environmental suitability) to 1 (high environmental suitability). We created separate ensemble maps for current and future scenarios (based on the ensemble of the three AOGCMs—CCSM4, MRI, and MIROC). To manage the raster datasets, we used the "raster" (Hijmans 2017), "rgdal" (Bivand et al. 2020), and "data.table" packages (Dowle and Srivnivasan 2020).

Effects of climate change on species redistribution

To test latitudinal, elevation (altitudinal), longitudinal, and slope shifts in species distribution over time, we converted final environmental suitability maps into presence-absence maps, considering values of > 10% quantile of suitability as presence and values of < 10% quantile as absence. Thus, we obtained the values of elevation (m), slope, latitude (decimal degrees), and longitude (decimal degrees) for each site predicted to be presence in each scenario (present, 2050 RCP 4.5, 2070 RCP 4.5, 2050 RCP 8.5, and 2070 RCP 8.5). Non-parametric Kruskal-Wallis tests followed by Dunn's post hoc test were performed to determine if dependent variables (elevation, slope, latitude, and longitude) differed between scenarios (present, 2050 RCP 4.5, 2070 RCP 4.5, 2050 RCP 8.5, and 2070 RCP 8.5). The magnitude of shifts in dependent variables over time were identified using the chi-square (χ^2) value provided by the Kruskal-Wallis test.

Elevation and slope values were obtained from <https://www.earthenv.org/DEM>. Kruskal-Wallis tests were performed using the "kruskal.test" function of the R stats package (R Core Team 2020), whereas Dunn's test was calculated using the "dunnTest" function of R, with Bonferroni p-values adjusted using the R FSA package (Ogle et al. 2020). Graphs were produced using the "ggplot" function from the ggplot2 package in R (Wickham 2016). All the analyses were performed using the R program (R Core Team 2020), considering a significance of $\alpha < 0.05$, and map projections were created using the QGIS program (QGIS Development Team 2020).

RESULTS

The ENMs for *Cercomacra melanaria* in the current scenario confirm that today's species distribution is restricted mainly to plains and wetlands, which are subject to the flood pulse (Fig. 3A). The ENMs for future scenarios indicate that the species' potential distribution will shift to higher lands and mountains (regions with higher slopes) in westward longitudes (Fig. 3B), but with a redistribution to the north (equatorial region) rather than the south (higher latitudes; Fig. 3B), as originally predicted (Fig. 2C). The model performance was considered adequate according to the TSS values (> 0.5; Appendix 3, Appendix 4).

Present and future distribution

The current predictions indicate that the species is distributed along the floodplains of the Brazilian Pantanal and Bolivian Chaco, in addition to the central Bolivian lowland regions comprising most of the Santa Cruz and Beni states (Fig. 3A). For RCP 4.5 scenarios (2050 and 2070), the species' habitat suitability should decrease in Brazilian and Bolivian Pantanal lowlands, and the sites of highest habitat suitability should shift toward highlands at the edges of the Upper Paraguay River basin in Brazil and northeastern Bolivia, and to the eastern side of the Andean highlands (Fig. 3B, Appendix 5, Appendix 6). The RCP 8.5 (2050 and 2070) prediction indicates that the species' habitat suitability should dwindle further and shift to highland regions. Areas with the highest habitat suitability are restricted to small highlands of the Paraguay basin in the Brazilian State of Mato Grosso do Sul, in extreme northwestern Bolivia, and small regions bordering the Andean highlands (Fig. 3B, Appendix 5, Appendix 6). Additionally, the species' potential area of distribution will decrease in future conditions, regardless of the climate scenario

(low or high greenhouse gas concentration), with a loss of approximately 92.97% (46,417,050 ha) in the most extreme scenario (RCP 8.5 2070—3,509,325 ha) compared to the current distribution (present—49,926,375 ha; Fig. 4A and Table 1).

Fig. 3. Potential habitat suitability and distribution map of the Mato Grosso Antbird (*Cercomacra melanaria*) in different climate scenarios (present, year 2050, year 2070) and different concentrations of greenhouse gases (Representative Concentration Pathway; RCP 4.5 and RCP 8.5). A = current species habitat suitability, B = climate scenarios. The colors indicate habitat suitability for the species.

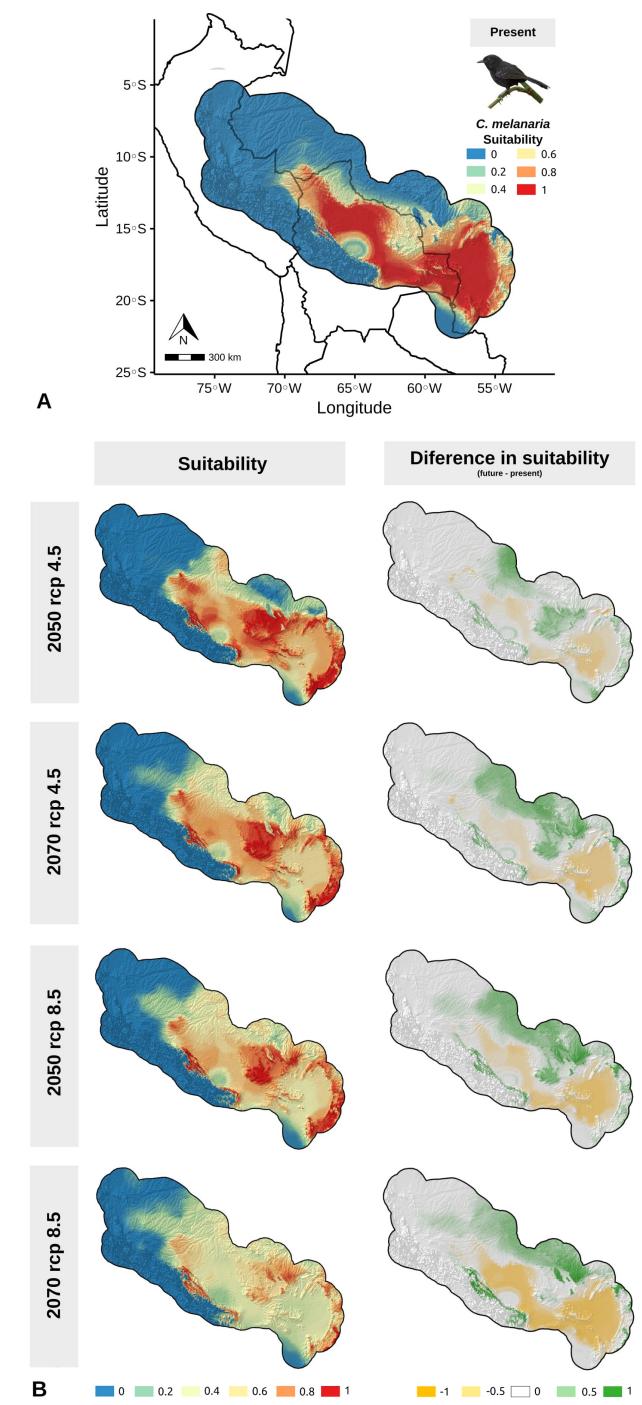


Table 1. Results of Mato Grosso Antbird (*Cercomacra melanaria*) potential distribution in the present and in future scenarios (Representative Concentration Pathway [RCP] 4.5 2050/2070 and RCP 8.5 2050/2070) of the different global circulation models (GCMs; CCSM4, MIROC, MRI) and mean of the three GCMs (Atmosphere-Ocean GCM's ensemble). The distribution area of the species is shown by square kilometers (km²), hectares (ha), and percentage (%).

Model	Area (km ²)	Area (ha)	%
Present potential distribution area	499 263.80	49 926 375	100
Models under RCP 4.5			
CSSM4 2050	391 695.80	39 169 575	78.46
CSSM4 2070	268 535.30	26 853 525	53.79
MIROC 2050	303 102.00	30 310 200	60.71
MIROC 2070	232 551.00	23 255 100	46.58
MRI 2050	302 474.30	30 247 425	60.59
MRI 2070	249 075.00	24 907 500	49.89
AOGCMs ensemble 2050	309 703.50	30 970 350	62.04
AOGCMs ensemble 2070	223 175.30	22 317 525	44.7
Models under RCP 8.5			
CSSM4 2050	221 109.80	22 110 975	44.29
CSSM4 2070	79 420.50	7 942 050	15.91
MIROC 2050	186 259.50	18 625 950	37.31
MIROC 2070	42 970.50	4 297 050	8.61
MRI 2050	182 027.30	18 202 725	36.46
MRI 2070	70 065.00	7 006 500	14.03
AOGCMs ensemble 2050	195 959.30	19 595 925	39.25
AOGCMs ensemble 2070	35 093.30	3 509 325	7.03

Effects of climate change on species redistribution

Future climate change will lead to significant shifts in species redistribution in terms of elevation ($\chi^2 = 25,697$, df = 4, p-value = < 0.001), slope ($\chi^2 = 17,024$, df = 4, p-value = < 0.001), latitude ($\chi^2 = 913.83$, df = 4, p-value = < 0.001) and longitude ($\chi^2 = 520.5$, df = 4, p-value = < 0.001; Fig. 4, Appendix 7). Based on chi-square values, it is clear that major shift occurs in altitude direction, i.e., the species distribution from lowlands to highlands, followed by slope (plains lands to mountainous land), latitude (south to north —equatorial region), and longitude (east to west).

Currently, species occur in lowlands with an elevation of 0 to 150 m, but the future climate scenarios show a gradual species' redistribution, mainly to highlands, ranging in elevation from 500 to 800 m (Fig. 4B). As for slope, the Mato Grosso Antbird today occurs in flat plains with low slopes (close to zero), while future climate scenarios indicate that the highest redistribution will occur toward areas with slopes of 0.5 to 2° (max: 5.79; min: 0.014; mean: 0.755; Fig. 4C). Latitudinal redistribution is predicted to shift the species toward lower latitudes (equatorial areas), with an average change from -17° (present) to -14° (future scenarios; Fig. 4D). Lastly, according to future scenarios, the Mato Grosso Antbird will tend to shift westward (to higher longitudes), with a mean change of -59° (present) to -62° (future scenarios). However, in the most contrasting scenario (RCP 8.5, 2070), the habitat suitability distribution is greater at longitudes of -67° and -55°, with an average of -61° (Fig. 4E; Appendix 8).

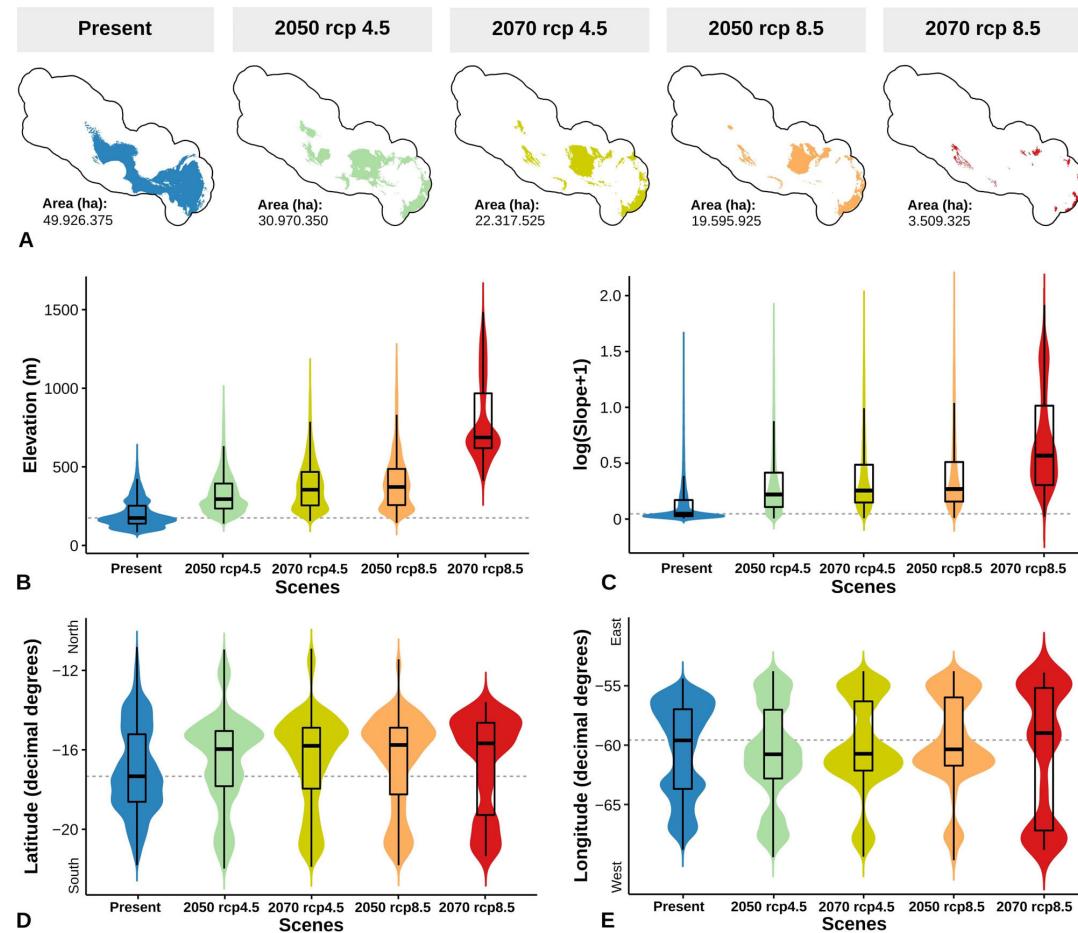
DISCUSSION

Our findings suggest a geographical shift of the Mato Grosso Antbird in response to future climate change. The species will suffer declining geographical distribution, with shifts in latitudinal, longitudinal, elevation, and slope patterns. The species currently occurs on plains and lowlands in South American wetlands. Its future northward and westward dispersion will cause the species to shift to highlands with steep slopes.

The future distribution scenarios suggest high species suitability at the headwaters of rivers in South American wetlands, mainly in Andean foothills (western Bolivia) and the border between Bolivia and Brazil (Fig. 4), corresponding to sites in Noel Kempff Mercado National Park (Bolivia) and the Serra Ricardo Franco State Park (Brazil). The upward shift to highlands will lead the species to water bodies and/or humid areas that match the current preference and tolerance of the species (Pinho et al. 2006, Bernardon et al. 2014).

Although the future predictions indicate upward shifts, colonization depends on the species' dispersal ability (e.g., forest corridors; Freeman et al. 2019). The Mato Grosso Antbird is a non-migratory and territorial species (Pinho et al. 2006) and, like most Thamnophilidae, has a low ability for long-distance flight (Zimmer et al. 2020). According to the hand-wing index (HWI), a morphological metric related to avian wing aspect ratio that is commonly used as a proxy of flight efficiency and dispersal ability (Sheard et al. 2020), the Mato Grosso Antbird's HWI (8.5) is considered low in comparison to that of other bird species (min

Fig. 4. Effects of future climate change across elevation, slope, latitude, and longitude for the Mato Grosso Antbird (*Cercomacra melanaria*) in the Pantanal. The dashed line indicates the median of present scenario. Slope is represented as log-transformed data to facilitate graph visualization. RCP = Representative Concentration Pathway.



= 0.016–max = 74.80; SD \pm 15.02). Hence, because of its low dispersal ability, the species will probably not move into its future suitable sites.

The move to higher lands may be facilitated by their proximity to the species' current distribution and by the existence of forest corridors, through which the species could hypothetically disperse more easily than by crossing open areas. Floodplain phytophysiognomies may also change in response to future global warming, increased climate seasonality, and/or anthropic disturbances such as deforestation. Guerra et al. (2020) infer a 3% loss in floodplain vegetation and a 10% loss in highland vegetation (where river headwaters are located) by 2050. Thus, the species dispersal range is expected to decline, considering future landscape configurations.

The future success of the species' geographical mobility will also depend on narrow-scale processes such as biological interactions, nesting sites, and resource availability. Nesting sites depend directly on vegetation types—forest vegetation in humid sites with dense understory (Bernardon et al. 2014). Precipitation rates are

important to maintain local humidity levels, but changes in precipitation seasonality and rising temperatures predicted for future global warming pose threats to floodplains (Enquist 2002, Li et al. 2009). Junk (2013) pointed out that precipitation rates in South America's floodplains will change dramatically in the future, with rainfall increasing in the rainy season and decreasing during the dry season, intensifying climate seasonality. Nesting depends mainly upon rainfall, given that the species' reproductive peak coincides with the onset of the rainy season because of higher availability of resources (invertebrates; Bernardon et al. 2014, Nóbrega et al. 2017). Thus, climate change also tends to directly affect the species' reproductive period.

Anthropogenic activities in and around river headwaters in the highlands represent another threat to the survival of the Mato Grosso Antbird. Pressure on the integrity of wetlands is increasing, mainly in the form of landscape modifications through mechanized agriculture and cattle raising, pollution, mining activities, and the construction of hydroelectric power plants (Junk 2013, Guerra et al. 2020, Medinas de Campos et al. 2020).

The construction of 104 hydroelectric plants in the basin that feeds the Pantanal is being proposed. If all these proposed dams were built, the area flooded by new dams would triple and close an additional 11,000 to 12,000 km of river channels (Medinas de Campos et al. 2020). One of the main ecological impacts of small hydropower plants (SHPs) is habitat loss (Medinas de Campos et al. 2020). The construction of these SHPs will affect mainly the water bodies and wetlands of the UPB plateau, corresponding to the most suitable locations for the Mato Grosso Antbird in the scenarios of future climate change.

The UPB plateau has also been subjected to landscape modifications by mechanized agriculture since the 1970s, highlighting the rapid and recent increase in land use conversion (SOS-Pantanal et al. 2017, Guerra et al. 2020). The expansion of monocultures in Pantanal highlands has caused soil erosion and silted up rivers, changing their flow and hydrological indices in lowlands, as has been observed in the Upper Taquari River basin (Assine 2005, Galdino et al. 2006). Loss of native vegetation through agriculture or hydropower facilities could diminish the suitability of these areas for the species, affecting the availability of food and nesting sites.

In the absence of quantitative data on population trends, a criterion for extinction risk assessment under future climate change is to quantify losses in species distribution area (km²). The IUCN Red List system is designed to account for uncertainty and allow the use of inference when appropriate (IUCN 2019). Forest loss data are also used to estimate extinction risk at local and regional scales. Forest loss is used as a proxy for population decline under criterion A of the Red List and calculation of area of occupation under criterion B (Tracewsky et al. 2016). According to Tracewsky et al. (2016), any species with a restricted home range and insufficient population data, such as the Mato Grosso Antbird, is considered at high risk of extinction, given its limited areas of occupation resulting from the paucity of remaining forest cover inside its range.

In view of the estimated future scenario for the distribution of the species, we have to consider that the current scenario of protected areas (PAs) may not be well balanced or representative. In Brazil, which is home to the largest range of the Mato Grosso Antbird, only 5.37% of the Pantanal is protected by Conservation Units, and this is one of the biomes least safeguarded by PAs in the country (MMA 2012, Chaves et al. 2018). The Pantanal is the world's largest continental wetland (Chardonnet et al. 2002, Harris et al. 2005), with four Ramsar sites (Ramsar Sites Information Service 2010). Nevertheless, the PAs do not cover the range needed to meet the area's biodiversity needs because 95% of the land in the Pantanal is privately owned and used mostly for ranching (Harris et al. 2005).

The threats are continuous. Between August and November 2020, 30% of this biome was destroyed by wildfires (Libonati et al. 2020), including protected areas and traditional communities. This was a destruction of unimaginable and catastrophic proportions, which revealed a scenario of extreme seasonality (longer dry season) that had already been predicted for the Pantanal region in climate change studies (Junk 2013, Marengo et al. 2015). Moreover, this tragedy also underscores the inability of the current government to manage threats within its own conservation areas (Ferrante and Fearnside 2019, Grasel et al.

2019), and the firefighting efforts were financed and carried out mostly by private individuals (M. F. Zuchetto, personal observation).

Although the current configuration of most PAs is not yet sufficiently representative, the sub-region of Chiquitano, Bolivia and the sub-Andean foothills were identified in a study as having some of the highest values of protected area coverage (17.3% and 11.7%, respectively, of extent considered as PAs; Prieto-Torres et al. 2018). This region coincides with some of the areas that we estimate as suitable in the future for the Mato Grosso Antbird. Therefore, there is an unequivocal need to increase the area and connectivity between PA networks to maximize long-term biodiversity conservation efforts (Prieto-Torres et al. 2018, Saura et al. 2018). Such an expansion would also connect the plains to areas of higher altitude, where most species are likely to find suitable areas considering a pessimistic scenario of climate change (Seimon et al. 2007, Şekercioğlu et al. 2008, 2012, Jankowski et al. 2013, Atauchi et al. 2020). Hence, the inclusion of the effects of climate change in the identification of priority areas for conservation helps to enhance the effectiveness of PA networks (Hannah et al. 2007; Triviño et al. 2018).

This study contributes by producing information about the distribution of a species with a restricted range, which is considered of least concern (LC), possibly because of data deficiency. This indicates that extinctions occur not only because of the disappearance of species classified as threatened, but also because of the disappearance of species with data deficiency. Therefore, in this paper we add knowledge to discuss the conservation status of the species.

CONCLUSION

The loss of ! 93% of the species' suitable distribution area (the most drastic future scenario, RCP 8.5 2070) and all the dangers we have listed here constitute a threatening scenario for the Mato Grosso Antbird. The scenario may become worse than predicted if one considers the fact that the species' low dispersal ability makes future suitable sites practically inaccessible. This makes it imperative to reassess the conservation status of the species, which is currently considered of LC.

The species response to future climate change depends on all the factors discussed above. Likewise, our findings also highlight the inherent risk of South American floodplain biodiversity in future climate scenarios.

*Responses to this article can be read online at:
<https://www.ace-eco.org/issues/responses.php/1893>*

Acknowledgments:

We are grateful for assistance and support from the National Institute of Science and Technology in Wetlands (INCT-INAU), the Higher Education Personnel Improvement Coordination (CAPES), and Mato Grosso State Research Support Foundation (FAPEMAT). We thank our collaborator Cristian Camilo Martinez González for the invaluable assistance with the

preliminary maps and data from Bolivia. Thanks also to the researchers Dennis Camacho, Miguel Angel Montenegro, Isabel Gomez, and Sofia Lothesberger for data collection from Bolivia and to Andressa Bach for reviewing the language of this manuscript.

LITERATURE CITED

- Allouche, O., A. Tsoar, and R. Kadmon. 2006. Assessing the accuracy of species distribution models: prevalence, kappa and the true skill statistic (TSS). *Journal of Applied Ecology* 43:1223-1232. <https://doi.org/10.1111/j.1365-2664.2006.01214.x>
- Araújo, M. B., R. P. Anderson, A. M. Barbosa, C. M. Beale, C. F. Dormann, R. Early, R. A. Garcia, A. Guisan, L. Maiorano, B. Naimi, R. B. O'Hara, N. E. Zimmermann, and C. Rahbek. 2019. Standards for distribution models in biodiversity assessments. *Science Advances* 5:1-12. <https://doi.org/10.1126/sciadv.aat4858>
- Araújo, M. B., and M. New. 2007. Ensemble forecasting of species distributions. *Trends in Ecology and Evolution* 22:42-47. <https://doi.org/10.1016/j.tree.2006.09.010>
- Assine, M. L. 2005. River avulsions on the Taquari megafan, Pantanal wetland, Brazil. *Geomorphology* 70:357-371. <https://doi.org/10.1016/j.geomorph.2005.02.013>
- Atauchi, P. J., C. Aucca-Chutas, G. Ferro, and D. A. Prieto-Torres. 2020. Present and future potential distribution of the endangered *Anairetes alpinus* (Passeriformes: Tyrannidae) under global climate change scenarios. *Journal of Ornithology* 161:723-738. <https://doi.org/10.1007/s10336-020-01762-z>
- Barve, N., V. Barve, A. Jiménez-Valverde, A. Lira-Noriega, S. P. Maher, A. T. Peterson, J. Soberón, and F. Villalobos. 2011. The crucial role of the accessible area in ecological niche modeling and species distribution modeling. *Ecological Modelling* 222:1810-1819. <https://doi.org/10.1016/j.ecolmodel.2011.02.011>
- Berg, M. P., E. T. Kiers, G. Driessen, M. van der Heijden, B. W. Kooi, F. Kuenen, M. Liefting, H. A. Verhoeft, and J. Ellers. 2010. Adapt or disperse: understanding species persistence in a changing world. *Global Change Biology* 16:587-598. <https://doi.org/10.1111/j.1365-2486.2009.02014.x>
- Bernardon, B., P. F. A. Nóbrega, and J. B. Pinho. 2014. Reproductive biology and nest-site selection of the Mato Grosso Antbird *Cercomacra melanaria* in the Brazilian Pantanal. *Revista Brasileira de Ornitologia* 22:270-277. [online] URL: http://www.revbrasilornitol.com.br/BJO/article/view/23_3_artigo5
- BirdLife International. 2018. *Cercomacra melanaria*. The IUCN Red List of Threatened Species 2018: e.T22701692A130215657. <https://dx.doi.org/10.2305/IUCN.UK.2018-2.RLTS.T22701692A130215657.en>
- Bivand, R., T. Keitt, and B. Rowlingson. 2020. rgdal: bindings for the 'geospatial' data abstraction library. R. package version 1.5-16. [online] URL: <https://CRAN.R-project.org/package=rgdal>
- Cardinale, B. J., J. E. Duffy, A. Gonzalez, D. U. Hooper, C. Perrings, P. Venail, A. Narwani, G. M. Mace, D. Tilman, D. A. Wardle, A. P. Kinzig, G. C. Daily, M. Loreau, J. B. Grace, A. Larigauderie, D. S. Srivastava, and S. Naeem. 2012. Biodiversity loss and its impact on humanity. *Nature* 486:59-67. <https://doi.org/10.1038/nature11148>
- Carpenter, G., A. N. Gillison, and J. Winter. 1993. DOMAIN: a flexible modelling procedure for mapping potential distributions of plants and animals. *Biodiversity and Conservation* 2:667-680. <https://doi.org/10.1007/BF00051966>
- Ceballos, G., P. R. Ehrlich, A. D. Barnosky, A. García, R. M. Pringle, and T. M. Palmer. 2015. Accelerated modern human-induced species losses: entering the sixth mass extinction. *Science Advances* 1(5). <https://doi.org/10.1126/sciadv.1400253>
- Chardonnet, Ph., B. des Clers, J. Fischer, R. Gerhold, F. Jori, and F. Lamarque. 2002. The value of wildlife. *Revue Scientifique et Technique* 21:15-51. <https://doi.org/10.20506/rst.21.1.1323>
- Chaves, J. V. B., and J. dos S. V. da Silva. 2018. Evolução das unidades de conservação no Pantanal no período de 1998 a 2018. Pages 678-685 in Anais 7º Simpósio de Geotecnologias no Pantanal. Jardim, Mato Grosso, Brazil. [online] URL <https://www.alice.cnptia.embrapa.br/alice/bitstream/doc/1099248/1/PLEvolucaoGeopantanal.pdf>
- Chen, I. C., H. J. Shiu, S. Benedick, J. D. Holloway, V. K. Chey, H. S. Barlow, J. K. Hill, and C. D. Thomas. 2009. Elevation increases in moth assemblages over 42 years on a tropical mountain. *Proceedings of the National Academy of Sciences of the United States of America* 106:1479-1483. <https://doi.org/10.1073/pnas.0809320106>
- Colwell, R. K., G. Brehm, C. L. Cardelús, A. C. Gilman, and J. T. Longino. 2008. Global warming, elevational range shifts, and lowland biotic attrition in the wet tropics. *Science* 322:258-261. <https://doi.org/10.1126/science.1162547>
- Crick, H. Q. P., S. R. Baillie, and D. I. Leech. 2003. The UK nest record scheme: its value for science and conservation. *Bird Study* 50:254-270. <https://doi.org/10.1080/00063650309461318>
- Crutzen P. J. 2006. The "Anthropocene". Pages 13-18 in E. Ehlers, and T. Krafft, editors. Earth system science in the anthropocene. Springer, Heidelberg, Germany. [online] URL <https://link.springer.com/content/pdf/10.1007%2Fb137853.pdf>
- de Piacentini, V. Q., A. Aleixo, C. E. Agne, G. N. Maurício, J. F. Pacheco, Gustavo A. Bravo, G. R. R. Brito, L. N. Naka, F. Olmos, S. Posso, L. F. Silveira, G. S. Betini, E. Carrano, I. Franz, A. C. Lees, L. M. Lima, D. Pioli, F. Schunck, F. R. do Amaral, G. A. Bencke, M. Cohn-Haft, L. F. A. Figueiredo, F. C. Straube, and E. Cesari. 2015. Annotated checklist of the birds of Brazil by the Brazilian Ornithological Records Committee. *Revista Brasileira de Ornitologia* 23:91-298. <https://doi.org/10.1007/BF03544294>
- Devictor, V., R. Julliard, and D. Couvet. 2008. Birds are tracking climate warming, but not fast enough. *Proceedings of the Royal Society B* 275:2743-2748. <https://doi.org/10.1098/rspb.2008.0878>
- Díaz-Almeyda, E. M., C. Prada, A. H. Ohdera, H. Moran, D. J. Civitello, R. Iglesias-Prieto, T. A. Carlo, T. C. Lajeunesse, and M. Medina. 2017. Intraspecific and interspecific variation in thermotolerance and photoacclimation in *Symbiodinium* dinoflagellates. *Proceedings of the Royal Society B* 284:20171767. <https://doi.org/10.1098/rspb.2017.1767>

- Diniz-Filho, J. A. F., L. M. Bini, T. F. Rangel, R. D. Loyola, C. Hof, D. Nogués-Bravo, and M. B. Araújo. 2009. Partitioning and mapping uncertainties in ensembles of forecasts of species turnover under climate change. *Ecography* 32:897-906. <https://doi.org/10.1111/j.1600-0587.2009.06196.x>
- Dirzo, R., and P. H. Raven. 2003. Global state of biodiversity and loss. *Annual Review of Environment and Resources* 28:137-167. <https://doi.org/10.1146/annurev.energy.28.050302.105532>
- Dirzo, R., H. S. Young, M. Galetti, G. Ceballos, N. J. B. Isaac, and B. Collen. 2014. Defaunation in the Anthropocene. *Science* 345:401-406. <https://doi.org/10.1126/science.1251817>
- Dowle, M., and A. Srinivasan. 2020. *data.table*: extension of ‘*data.frame*.’ R. package 1.13.0. [online] URL: <https://CRAN.R-project.org/package=data.table>
- Enquist, C. A. F. 2002. Predicted regional impacts of climate change on the geographical distribution and diversity of tropical forests in Costa Rica. *Journal of Biogeography* 29:519-534. <https://doi.org/10.1046/j.1365-2699.2002.00695.x>
- Fei, S., J. M. Desprez, K. M. Potter, I. Jo, J. A. Knott, and C. M. Oswalt. 2017. Divergence of species responses to climate change. *Science Advances* 3:e1603055. <https://doi.org/10.1126/sciadv.1603055>
- Ferrante, L., and P. M. Fearnside. 2019. Brazil’s new president and “ruralists” threaten Amazonia’s environment, traditional peoples and the global climate. *Environmental Conservation* 46:261-263. <https://doi.org/10.1017/S0376892919000213>
- Francou, B., M. Vuille, P. Wagnon, J. Mendoza, and J. E. Sicart. 2003. Tropical climate change recorded by a glacier in the central Andes during the last decades of the twentieth century: Chacaltaya, Bolivia, 16°S. *Journal of Geophysical Research* 108:1-12. <https://doi.org/10.1029/2002jd002959>
- Franklin, J. 2010. Mapping species distributions: spatial inference and prediction. Cambridge University Press, Cambridge, UK. <https://doi.org/10.1017/CBO9780511810602>
- Freeman, B., J. Sunnarborg, and A. T. Peterson. 2019. Effects of climate change on the distributional potential of three range-restricted West African bird species. *Condor* 121:1-10. <https://doi.org/10.1093/condor/duz012>
- Galdino, S., L. M. Vieira, and L. A. Pellegrin. 2006. Environmental and socioeconomic impacts in the Taquari River Basin - Pantanal. Embrapa Pantanal, Corumbá, Mato Grosso do Sul, Brazil. [online] URL: <https://ainfo.cnptia.embrapa.br/digital/bitstream/item/38652/1/Livro025.pdf>
- GBIF.org. 2019. GBIF occurrence download. Global Biodiversity Information Facility, Copenhagen, Denmark. <https://doi.org/10.15468/dl.zr39py>
- Graham, C. H., N. Silva, and J. Velásquez-Tibatá. 2010. Evaluating the potential causes of range limits of birds of the Colombian Andes. *Journal of Biogeography* 37:1863-1875. <https://doi.org/10.1111/j.1365-2699.2010.02356.x>
- Grasel, D., P. M. Fearnside, A. S. Rovai, J. R. S. Vitule, R. R. Rodrigues, R. P. Mormul, F. D. F. Sampaio, and J. A. Jarenkow. 2019. Brazil’s Native Vegetation Protection Law jeopardizes wetland conservation: a comment on Maltchik et al. *Environmental Conservation* 46:121-123. <https://doi.org/10.1017/S0376892918000474>
- Guerra, A., F. de O. Roque, L. C. Garcia, J. M. Ochoa-Quintero, P. T. S. de Oliveira, R. D. Guariento, and I. M. D. Rosa. 2020. Drivers and projections of vegetation loss in the Pantanal and surrounding ecosystems. *Land Use Policy* 91:104388. <https://doi.org/10.1016/j.landusepol.2019.104388>
- Hannah, L., G. Midgley, S. Andelman, M. Araújo, G. Hughes, E. Martinez-Meyer, R. Pearson, and P. Williams. 2007. Protected area needs in a changing climate. *Frontiers in Ecology and the Environment* 5:131-138. [online] URL: <https://www.jstor.org/stable/20440608>
- Hao, T., J. Elith, G. Guillera-Arroita, and J. J. Lahoz-Monfort. 2019. A review of evidence about use and performance of species distribution modelling ensembles like BIOMOD. *Diversity and Distributions* 25:839-852. <https://doi.org/10.1111/ddi.12892>
- Harris, M. B., W. Tomas, G. Mourão, C. J. Da Silva, E. Guimarães, F. Sonoda, and E. Fachim. 2005. Safeguarding the Pantanal wetlands: threats and conservation initiatives. *Conservation Biology* 19:714-720. <https://doi.org/10.1111/j.1523-1739.2005.00708.x>
- Hijmans, R. J. 2017. *Raster*: geographic data analysis and modeling. [online] URL: <http://CRAN.R-project.org/package=raster>
- Hijmans, R. J., S. E. Cameron, J. L. Parra, P. G. Jones, and A. Jarvis. 2005. Very high resolution interpolated climate surfaces for global land areas. *International Journal of Climatology* 25:1965-1978. <https://doi.org/10.1002/joc.1276>
- Hijmans, R. J., S. Phillips, J. Leathwick, and J. Elith. 2015. *dismo*: species distribution modeling. [online] URL: <https://CRAN.R-project.org/package=dismo>
- INaturalist.org. 2019. Mato Grosso Antbird *Cercomacra melanaria*. [online] URL: <https://www.inaturalist.org/taxa/15737-Cercomacra-melanaria>
- Intergovernmental Panel on Climate Change (IPCC). 2014. Climate change 2014: synthesis report. Contribution of Working Groups I, II and III to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change. Core Writing Team, R. K. Pachauri, and L. A. Meyer, editors. IPCC, Geneva, Switzerland. [online] URL: <https://epic.awi.de/id/eprint/37530/>
- International Union for Conservation of Nature (IUCN). Standards and Petitions Subcommittee. 2019. Guidelines for using the IUCN Red List categories and criteria. Version 14. [online] URL: <http://www.iucnredlist.org/documents/RedListGuidelines.pdf>
- Jankowski, J. E., C. L. Merkord, W. F. Rios, K. G. Cabrera, N. S. Revilla, and M. R. Silman. 2013. The relationship of tropical bird communities to tree species composition and vegetation structure along an Andean elevational gradient. *Journal of Biogeography* 40:950-962. <https://doi.org/10.1111/jbi.12041>
- Junk, W. J. 2013. Current state of knowledge regarding South America wetlands and their future under global climate change. *Aquatic Sciences* 75:113-131. <https://doi.org/10.1007/s00027-012-0253-8>

- Karatzoglou, A., A. Smola, K. Hornik, and A. Zeileis. 2004. kernlab – an S4 package for kernel methods in R. *Journal of Statistical Software* 11:1-20. <https://doi.org/10.18637/jss.v011.i09>
- Kelly, A. E., and M. L. Goulden. 2008. Rapid shifts in plant distribution with recent climate change. *Proceedings of the National Academy of Sciences of the United States of America* 105:11823-11826. <https://doi.org/10.1073/pnas.0802891105>
- Kinzelbach, R. K. 1995. Vogelwelt und klimaveränderung im 16. jahrhundert. *Naturwissenschaften* 82: 499–508. <https://doi.org/10.1007/BF01134485>
- Lee, A. T. K., and H. Nel. 2020. BirdLasser: the influence of a mobile app on a citizen science project. *African Zoology* 55:155-160. <https://doi.org/10.1080/15627020.2020.1717376>
- Li, J., D. W. Hilbert, T. Parker, and S. Williams. 2009. How do species respond to climate change along an elevation gradient? A case study of the grey-headed robin (*Heteromyias albисpecularis*). *Global Change Biology* 15:255-267. <https://doi.org/10.1111/j.1365-2486.2008.01737.x>
- Libonati, R., J. A. Rodrigues, and F. Lemos. 2020. Nota técnica LASA – Área queimada Pantanal: situação até 15 de novembro 2020. Laboratório de Aplicações de Satélites Ambientais, Departamento de Meteorologia, Rio de Janeiro, Brazil. <https://doi.org/10.13140/RG.2.2.12810.90564>
- Liu, C., G. Newell, and M. White. 2016. On the selection of thresholds for predicting species occurrence with presence-only data. *Ecology and Evolution* 6:337-348. <https://doi.org/10.1002/ece3.1878>
- Marengo, J. A., G. S. Oliveira, and L. M. Alves. 2015. Climate change scenarios in the Pantanal. Pages 227-238 in I. Bergier and M. L. Assine, editors. *Dynamics of the Pantanal Wetland in South America*. Springer International, Switzerland. https://doi.org/10.1007/978-3-319-2015_357
- Medinas de Campos, M., H. M. Tritico, P. Girard, P. Zeilhofer, S. K. Hamilton, and I. Fantin-Cruz. 2020. Predicted impacts of proposed hydroelectric facilities on fish migration routes upstream from the Pantanal wetland (Brazil). *River Research and Applications* 36:452-464. <https://doi.org/10.1002/rra.3588>
- Ménétrier, E. 1835. Monographie de la famille des Myotherinae où sont décrites les espèces qui ornent le Musée d'Académie impériale des Sciences. *Mémoires de l'Académie des Sciences de St. Pétersburg*, 6th serie 3:443–544. [online] URL: <https://www.biodiversitylibrary.org/page/28731626>
- Ministério do Meio Ambiente (MMA). 2012. Cadastro Nacional de Unidades de Conservação. Brasília, Brasil. [online] URL: <https://antigo.mma.gov.br/areas-protegidas/cadastro-nacional-de-ucs/itemlist/category/130-cadastro-nacional-de-uc-s.html>
- Morrone, J. J. 2014. Biogeographical regionalisation of the neotropical region. *Zootaxa* 3782:1-110. <https://doi.org/10.11646/zootaxa.3782.1.1>
- Nix, H.A. 1986. A biogeographic analysis of Australian elapid snakes. Pages 4-15 in R. Longmore, editor. *Atlas of elapid snakes of Australia*. Australian Government Publishing Service, Canberra, Australia.
- Nóbrega, P. F. A., J. B. de Pinho, and C. Duca. 2017. Population viability analysis of the Mato Grosso Antbird (*Cercomacra melanaria*) in the Pantanal of Mato Grosso, Brazil. *Revista Brasileira de Ornitologia* 25:169-175. <https://doi.org/10.1007/BF03544395>
- Nunes, A. P. 2011. Quantas espécies de aves ocorrem no Pantanal brasileiro? Atualidades Ornitológicas. 160:45-54. [online] URL: http://www.ao.com.br/download/AO160_45.pdf
- Ogle, D.H., J. C. Doll, P. Wheeler, and A. Dinno. 2020. FSA: Fisheries stock analysis. R package version 0.8.30. [online] URL: <https://github.com/droglenc/FSA>
- Parmesan, C., and G. Yohe. 2003. A globally coherent fingerprint of climate change. *Nature* 421:37-42. <https://doi.org/10.1038/nature01286>
- Pearce-Higgins, J. W., and R. E. Green. 2014. Birds and climate change: impacts and conservation responses. Cambridge University Press, Cambridge, UK. <https://doi.org/10.1017/CBO9781139047791>
- Pecl, G. T., M. B. Araújo, J. D. Bell, J. Blanchard, T. C. Bonebrake, I. Chen, T. D. Clark, R. K. Colwell, F. Danielsen, B. Evengård, L. Falconi, S. Ferrier, S. Frusher, R. A. Garcia, R. B. Griffis, A. J. Hobday, C. Janion-Scheepers, M. A. Jarzyna, S. Jennings, J. Lenoir, H. I. Linnetved, V. Y. Martin, P. C. McCormack, J. McDonald, N. J. Mitchell, T. Mustonen, J. M. Pandolfi, N. Pettorelli, E. Popova, S. A. Robinson, B. R. Scheffers, J. D. Shaw, C. J. B. Sorte, J. M. Strugnell, J. M. Sunday, M. N. Tuanmu, A. Vergés, C. Villanueva, T. Wernberg, E. Wapstra, and S. E. Williams. 2017. Biodiversity redistribution under climate change: impacts on ecosystems and human well-being. *Science* 355:1-9. <https://doi.org/10.1126/science.aai9214>
- Peterson, A. T., M. A. Ortega-Huerta, J. Bartley, V. Sánchez-Cordero, J. Soberón, R. H. Buddeleher, and D. R. B. Stockwell. 2002. Future projections for Mexican faunas under global climate change scenarios. *Nature* 416:626-629. <https://doi.org/10.1038/416626a>
- Peterson, A. T., J. Soberón, R. G. Pearson, R. P. Anderson, E. Martínez-Meyer, M. Nakamura, and M.B. Araújo. 2011. Ecological niches and geographic distributions. Princeton University Press, Princeton, New Jersey, USA.
- Phillips, S. J., and M. Dudík. 2008. Modeling of species distributions with Maxent: new extensions and a comprehensive evaluation. *Ecography* 31:161-175. <https://doi.org/10.1111/j.0906-7590.2008.5203.x>
- Pinho, J. B., L. E. Lopes, D. H. de Moraes, and A. M. Fernandes. 2006. Life history of the Mato Grosso Antbird *Cercomacra melanaria* in the Brazilian Pantanal. *Ibis* 148:321-329. <https://doi.org/10.1111/j.1474-919X.2006.00542.x>
- Prieto-Torres, D. A., A. Lira-Noriega, and A. G. Navarro-Sigüenza. 2020. Climate change promotes species loss and uneven modification of richness patterns in the avifauna associated to Neotropical seasonally dry forests. *Perspectives in Ecology and Conservation* 18:19-30. <https://doi.org/10.1016/j.pecon.2020.01.002>
- Prieto-Torres, D. A., J. Nori, and O. R. Rojas-Soto. 2018. Identifying priority conservation areas for birds associated to

- endangered Neotropical dry forests. *Biological Conservation* 228:205-214. <https://doi.org/10.1016/j.biocon.2018.10.025>
- QGIS Development Team. 2020. QGIS geographic information system. Open Source Geospatial Foundation Project. [online] URL: <http://qgis.osgeo.org>
- R Core Team. 2020. R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria. [online] URL: <https://www.R-project.org/>
- Ramsar Sites Information Service. 2010. Annotated list of wetlands of international importance, Brazil. [online] URL: https://rsis.ramsar.org/sites/default/files/rsiswp_search/exports/Ramsar-Sites-annotated-summary-Brazil.pdf?1611261955
- Revelle, W. 2020. psych: Procedures for personality and psychological research, version 2.0.9. Northwestern University, Evanston, Illinois, USA. [online] URL: <https://CRAN.R-project.org/package=psych>.
- Root, T. L., J. T. Price, K. R. Hall, S. H. Schneider, C. Rosenzweig, and J. A. Pounds. 2003. Fingerprints of global warming on wild animals and plants. *Nature* 421:57-60. <https://doi.org/10.1038/nature01333>
- Santillán, V., M. Quijón, B. A. Tinoco, E. Zárate, M. Schleuning, K. Böhning-Gaese, and E. L. Neuschulz. 2020. Direct and indirect effects of elevation, climate and vegetation structure on bird communities on a tropical mountain. *Acta Oecologica* 102:103500. <https://doi.org/10.1016/j.actao.2019.103500>
- Saura, S., B. Bartzky, L. Bastin, L. Battistella, A. Mandriči, and G. Dubois. 2018. Protected area connectivity: shortfalls in global targets and country-level priorities. *Biological Conservation* 219:53-67. <https://doi.org/10.1016/j.biocon.2017.12.020>
- Seimon, T. A., A. Seimon, P. Daszak, S. R. P. Halloy, L. M. Schloegel, C. A. Aguilar, P. Sowell, A. D. Hyatt, B. Konecky, and J. E. Simmons. 2007. Upward range extension of Andean anurans and chytridiomycosis to extreme elevations in response to tropical deglaciation. *Global Change Biology* 13:288-299. <https://doi.org/10.1111/j.1365-2486.2006.01278.x>
- Şekercioğlu, C. H. 2006. Increasing awareness of avian ecological function. *Trends in Ecology and Evolution* 21:464-471. <https://doi.org/10.1016/j.tree.2006.05.007>
- Şekercioğlu, C. H., R. B. Primack, and J. Wormworth. 2012. The effects of climate change on tropical birds. *Biological Conservation* 148:1-18. <https://doi.org/10.1016/j.biocon.2011.10.019>
- Şekercioğlu, C. H., S. H. Schneider, J. P. Fay, and S. R. Loarie. 2008. Climate change, elevational range shifts, and bird extinctions. *Conservation Biology* 22:140-150. <https://doi.org/10.1111/j.1523-1739.2007.00852.x>
- Sheard, C., M. H. C. Neate-Clegg, N. Alioravainen, S. E. I. Jones, C. Vincent, H. E. A. Macgregor, T. P. Bregman, S. Claramunt, and J. A. Tobias. 2020. Ecological drivers of global gradients in avian dispersal inferred from wing morphology. *Nature Communications* 11:2463. <https://doi.org/10.1038/s41467-020-16313-6>
- Sobral-souza, T., R. B. Francini, and M. S. Lima-Ribeiro. 2015. Species extinction risk might increase out of reserves: allowances for conservation of threatened butterfly *Actinote quadra* (Lepidoptera: Nymphalidae) under global warming. *Natureza & Conservação* 13:159-165. <https://doi.org/10.1016/j.ncon.2015.11.009>
- SOS-Pantanal, WWF-Brasil, Conservation-International, ECOA, Fundación-AVINA, (2017). Monitoramento das alterações da cobertura vegetal e uso do solo na Bacia Do Alto Paraguai Porção Brasileira—período de análise: 2016 a 2017. Embrapa Pantanal, Corumbá, Mato Grosso do Sul, Brasil. [online] URL: <https://www.embrapa.br/documents/1354999/1529097/BAP++Mapping+da+Bacia+do+Alto+Paraguay+-+sum%C3%A1rio+executivo/a2312a90-c14f-4f39-b43d-61d6dea43556>
- Stigall, A. L. 2012. Using ecological niche modelling to evaluate niche stability in deep time. *Journal of Biogeography* 39:772-781. <https://doi.org/10.1111/j.1365-2699.2011.02651.x>
- Sullivan, B. L., J. L. Aycrigg, J. H. Barry, R. E. Bonney, N. Bruns, C. B. Cooper, T. Damoulas, A. A. Dhondt, T. Dietterich, A. Farnsworth, D. Fink, J. W. Fitzpatrick, T. Fredericks, J. Gerbracht, C. Gomes, W. M. Hochachka, M. J. Iliff, C. Lagoze, F. A. La Sorte, M. Merrifield, W. Morris, T. B. Phillips, M. Reynolds, A. D. Rodewald, K. V. Rosenberg, N. M. Trautmann, A. Wiggins, D. W. Winkler, W. Wong, C. L. Wood, J. Yu, and S. Kelling. 2014. The eBird enterprise: an integrated approach to development and application of citizen science. *Biological Conservation* 169:31-40. <https://doi.org/10.1016/j.biocon.2013.11.003>
- Sunday, J. M., A. E. Bates, and N. K. Dulvy. 2012. Thermal tolerance and the global redistribution of animals. *Nature Climate Change* 2:686-690. <https://doi.org/10.1038/nclimate1539>
- Tax, D. M. J., and R. P. W. Duin. 2004. Support vector data description. *Machine Learning* 54:45-66. <https://doi.org/10.1023/B:MACH.0000008084.60811.49>
- Tracewski, Ł., S. H. M. Butchart, M. Di Marco, G. F. Ficetola, C. Rondinini, A. Symes, H. Wheatley, A. E. Beresford, and G. M. Buchanan. 2016. Toward quantification of the impact of 21st-century deforestation on the extinction risk of terrestrial vertebrates. *Conservation Biology* 30:1070-1079. <https://doi.org/10.1111/cobi.12715>
- Triviño, M., H. Kujala, M. B. Araújo, and M. Cabeza. 2018. Planning for the future: identifying conservation priority areas for Iberian birds under climate change. *Landscape Ecology* 33:659-673. <https://doi.org/10.1007/s10980-018-0626-z>
- Walsberg, G. E. 1993. Thermal consequences of diurnal microhabitat selection in a small bird. *Ornis Scandinavica* 24:174-182. <https://doi.org/10.2307/3676733>
- Wickham H. 2016. ggplot2: elegant graphics for data analysis. Springer-Verlag, New York, New York, USA.
- Wormworth, J., and C.H. Şekercioğlu. 2011. Winged sentinels: birds and climate change. Cambridge University Press, Melbourne, AU. <https://doi.org/10.1017/CBO9781139150026>

Zimmer, K., M. L. Isler, and E. de Juana. 2020. Mato Grosso Antbird (*Cercomacra melanaria*), version 1.0. In J. del Hoyo, A. Elliott, J. Sargatal, D. A. Christie, and E. de Juana, editors. Birds of the world. Cornell Lab of Ornithology, Ithaca, New York, USA. <https://doi.org/10.2173/bow.magant1.01>

Editor-in-Chief: Alexander L. Bond
Subject Editor: Sergio A. Cabrera Cruz



Sponsored by the Society of
Canadian Ornithologists and
Birds Canada

Parrainée par la Société des
ornithologues du Canada et
Oiseaux Canada



Appendix 1

Table A1.1: Occurrence data from citizen science databases, digital databases (Global Biodiversity Information Facility - GBIF, Museo Nacional de Historia Natural, Colección Boliviana de Ornitológia, Universidad Mayor de San Andrés -UMSA) and directly with researchers (Dennis Camacho - Asociación Boliviana de Ornitológia – ASBOR, Miguel Angel Montenegro - Museo de Historia Natural Noel Kempff Mercado/ Universidad Autónoma Gabriel René Moreno - UAGRM, Sofia Lothesberger and Isabel Gomez-UMSA).

Name	Longitude	Latitude	Year	Source
Cercomacra_melanaria	-56.364	-20.243	2019	gbif
Cercomacra_melanaria	-56.876	-16.758	2019	gbif
Cercomacra_melanaria	-56.304	-19.953	2018	gbif
Cercomacra_melanaria	-58.161	-20.078	2018	gbif
Cercomacra_melanaria	-64.754	-15.193	2018	gbif
Cercomacra_melanaria	-56.409	-20.225	2018	gbif
Cercomacra_melanaria	-55.77	-16.132	2018	gbif
Cercomacra_melanaria	-57.039	-19.556	2018	gbif
Cercomacra_melanaria	-56.408	-20.256	2018	gbif
Cercomacra_melanaria	-55.791	-20.478	2018	gbif
Cercomacra_melanaria	-55.965	-20.096	2018	gbif
Cercomacra_melanaria	-55.823	-16.344	2018	gbif
Cercomacra_melanaria	-56.417	-14.833	2018	gbif
Cercomacra_melanaria	-56.775	-17.365	2018	gbif
Cercomacra_melanaria	-56.621	-16.379	2018	gbif
Cercomacra_melanaria	-64.746	-14.997	2018	gbif
Cercomacra_melanaria	-56.699	-17.283	2018	gbif
Cercomacra_melanaria	-56.745	-16.503	2018	gbif
Cercomacra_melanaria	-56.735	-16.621	2018	gbif
Cercomacra_melanaria	-56.718	-17.315	2018	gbif
Cercomacra_melanaria	-56.943	-17.103	2018	gbif
Cercomacra_melanaria	-56.693	-17.322	2018	gbif
Cercomacra_melanaria	-56.769	-17.355	2018	gbif
Cercomacra_melanaria	-56.832	-17.409	2018	gbif
Cercomacra_melanaria	-56.844	-16.784	2018	gbif
Cercomacra_melanaria	-56.712	-16.512	2018	gbif
Cercomacra_melanaria	-56.24	-15.372	2018	gbif
Cercomacra_melanaria	-56.427	-20.241	2018	gbif
Cercomacra_melanaria	-56.823	-17.458	2018	gbif
Cercomacra_melanaria	-56.753	-16.598	2018	gbif

Cercomacra_melanaria	-56.838	-17.299	2018	gbif
Cercomacra_melanaria	-65.051	-14.005	2018	gbif
Cercomacra_melanaria	-56.386	-16.525	2018	gbif
Cercomacra_melanaria	-56.577	-17.168	2018	gbif
Cercomacra_melanaria	-56.595	-17.331	2018	gbif
Cercomacra_melanaria	-56.615	-20.086	2018	gbif
Cercomacra_melanaria	-64.856	-14.877	2018	gbif
Cercomacra_melanaria	-64.907	-14.74	2018	gbif
Cercomacra_melanaria	-64.974	-14.796	2018	gbif
Cercomacra_melanaria	-55.944	-20.141	2018	gbif
Cercomacra_melanaria	-64.813	-14.911	2018	gbif
Cercomacra_melanaria	-56.399	-16.472	2018	gbif
Cercomacra_melanaria	-56.641	-16.562	2018	gbif
Cercomacra_melanaria	-57.04	-19.401	2018	gbif
Cercomacra_melanaria	-64.943	-14.206	2018	gbif
Cercomacra_melanaria	-64.889	-14.849	2018	gbif
Cercomacra_melanaria	-56.873	-20.737	2018	gbif
Cercomacra_melanaria	-57.038	-19.575	2018	gbif
Cercomacra_melanaria	-54.926	-20.227	2018	gbif
Cercomacra_melanaria	-56.729	-16.534	2018	gbif
Cercomacra_melanaria	-56.594	-16.404	2018	gbif
Cercomacra_melanaria	-56.551	-17.414	2018	gbif
Cercomacra_melanaria	-56.59	-16.396	2018	gbif
Cercomacra_melanaria	-57.602	-19.024	2018	gbif
Cercomacra_melanaria	-57.363	-18.458	2018	gbif
Cercomacra_melanaria	-64.937	-14.81	2018	gbif
Cercomacra_melanaria	-57.382	-18.39	2018	gbif
Cercomacra_melanaria	-57.384	-18.168	2018	gbif
Cercomacra_melanaria	-57.492	-18.078	2018	gbif
Cercomacra_melanaria	-57.655	-18.997	2018	gbif
Cercomacra_melanaria	-56.138	-15.545	2017	gbif
Cercomacra_melanaria	-56.691	-16.457	2017	gbif
Cercomacra_melanaria	-56.736	-17.312	2017	gbif
Cercomacra_melanaria	-56.674	-16.41	2017	gbif
Cercomacra_melanaria	-64.479	-13.553	2017	gbif
Cercomacra_melanaria	-56.929	-17.117	2017	gbif
Cercomacra_melanaria	-56.706	-16.61	2017	gbif
Cercomacra_melanaria	-56.675	-16.504	2017	gbif
Cercomacra_melanaria	-56.378	-16.513	2017	gbif
Cercomacra_melanaria	-57.837	-16.567	2017	gbif
Cercomacra_melanaria	-56.674	-17.335	2017	gbif

Cercomacra_melanaria	-56.713	-20.536	2017	gbif
Cercomacra_melanaria	-56.298	-16.367	2017	gbif
Cercomacra_melanaria	-63.07	-17.766	2017	gbif
Cercomacra_melanaria	-58.204	-19.966	2017	gbif
Cercomacra_melanaria	-56.648	-16.362	2017	gbif
Cercomacra_melanaria	-56.674	-16.453	2016	gbif
Cercomacra_melanaria	-55.613	-19.507	2016	gbif
Cercomacra_melanaria	-56.624	-16.273	2016	gbif
Cercomacra_melanaria	-56.669	-16.592	2016	gbif
Cercomacra_melanaria	-55.39	-19.285	2016	gbif
Cercomacra_melanaria	-56.32	-15.722	2016	gbif
Cercomacra_melanaria	-56.837	-16.848	2016	gbif
Cercomacra_melanaria	-56.681	-17.268	2016	gbif
Cercomacra_melanaria	-64.993	-14.224	2016	gbif
Cercomacra_melanaria	-56.771	-16.644	2016	gbif
Cercomacra_melanaria	-65.015	-14.209	2016	gbif
Cercomacra_melanaria	-56.91	-17.25	2016	gbif
Cercomacra_melanaria	-57.516	-19.203	2016	gbif
Cercomacra_melanaria	-56.575	-20.225	2015	gbif
Cercomacra_melanaria	-57.844	-21.1	2015	gbif
Cercomacra_melanaria	-58.133	-20.383	2015	gbif
Cercomacra_melanaria	-64.74	-15.153	2015	gbif
Cercomacra_melanaria	-64.891	-14.892	2015	gbif
Cercomacra_melanaria	-56.141	-15.755	2015	gbif
Cercomacra_melanaria	-65.213	-13.693	2015	gbif
Cercomacra_melanaria	-56.641	-16.33	2015	gbif
Cercomacra_melanaria	-56.714	-16.527	2014	gbif
Cercomacra_melanaria	-56.35	-16.381	2014	gbif
Cercomacra_melanaria	-56.627	-16.267	2014	gbif
Cercomacra_melanaria	-56.088	-19.589	2014	gbif
Cercomacra_melanaria	-57.017	-19.267	2014	gbif
Cercomacra_melanaria	-57.231	-19.255	2013	gbif
Cercomacra_melanaria	-65.929	-14.941	2013	gbif
Cercomacra_melanaria	-58.146	-20.134	2013	gbif
Cercomacra_melanaria	-56.842	-16.701	2013	gbif
Cercomacra_melanaria	-55.863	-16.346	2013	gbif
Cercomacra_melanaria	-57.787	-18.969	2012	gbif
Cercomacra_melanaria	-56.786	-17.004	2012	gbif
Cercomacra_melanaria	-64.97	-14.868	2012	gbif
Cercomacra_melanaria	-56.602	-19.668	2012	gbif
Cercomacra_melanaria	-55.767	-19.829	2012	gbif

Cercomacra_melanaria	-64.76	-15.192	2012	gbif
Cercomacra_melanaria	-56.066	-15.635	2012	gbif
Cercomacra_melanaria	-64.975	-14.462	2011	gbif
Cercomacra_melanaria	-56.76	-16.758	2011	gbif
Cercomacra_melanaria	-64.946	-14.725	2011	gbif
Cercomacra_melanaria	-64.929	-14.113	2011	gbif
Cercomacra_melanaria	-56.765	-16.557	2010	gbif
Cercomacra_melanaria	-56.957	-19.494	2010	gbif
Cercomacra_melanaria	-63.736	-17.002	2010	gbif
Cercomacra_melanaria	-56.151	-19.577	2010	gbif
Cercomacra_melanaria	-64.958	-14.748	2010	gbif
Cercomacra_melanaria	-64.77	-14.684	2010	gbif
Cercomacra_melanaria	-58.269	-20.184	2009	gbif
Cercomacra_melanaria	-55.507	-20.459	2009	gbif
Cercomacra_melanaria	-65.113	-13.643	2009	gbif
Cercomacra_melanaria	-64.913	-14.003	2008	gbif
Cercomacra_melanaria	-63.07	-17.774	2008	gbif
Cercomacra_melanaria	-58.638	-17.26	2008	gbif
Cercomacra_melanaria	-64.768	-15.266	2007	gbif
Cercomacra_melanaria	-66.869	-14.214	2007	gbif
Cercomacra_melanaria	-56.854	-15.169	2007	gbif
Cercomacra_melanaria	-58.2	-19.983	2006	gbif
Cercomacra_melanaria	-58.17	-20.23	2006	gbif
Cercomacra_melanaria	-56.426	-17.665	2006	gbif
Cercomacra_melanaria	-56.667	-21.009	2006	gbif
Cercomacra_melanaria	-66.824	-14.073	2006	gbif
Cercomacra_melanaria	-57.87	-21.036	2006	gbif
Cercomacra_melanaria	-64.509	-13.805	2006	gbif
Cercomacra_melanaria	-63.829	-13.232	2005	gbif
Cercomacra_melanaria	-56.833	-16.667	2005	gbif
Cercomacra_melanaria	-56.22	-20.292	2004	gbif
Cercomacra_melanaria	-55.893	-15.696	2003	gbif
Cercomacra_melanaria	-56.803	-17.324	2003	gbif
Cercomacra_melanaria	-55.677	-19.777	2003	gbif
Cercomacra_melanaria	-57.385	-19.031	2002	gbif
Cercomacra_melanaria	-56.47	-16.372	2001	gbif
Cercomacra_melanaria	-62.509	-19.175	2001	gbif
Cercomacra_melanaria	-62.521	-18.959	2001	gbif
Cercomacra_melanaria	-61.959	-18.391	2000	gbif
Cercomacra_melanaria	-56.845	-16.825	2000	gbif
Cercomacra_melanaria	-58.3	-19.92	1999	gbif

Cercomacra_melanaria	-58.133	-20.1	1999	gbif
Cercomacra_melanaria	-64.981	-13.915	1998	gbif
Cercomacra_melanaria	-60.31	-20.31	1998	gbif
Cercomacra_melanaria	-66.288	-14.645	1998	gbif
Cercomacra_melanaria	-58.417	-20.05	1997	gbif
Cercomacra_melanaria	-64.611	-13.264	1994	gbif
Cercomacra_melanaria	-63.161	-17.948	1992	gbif
Cercomacra_melanaria	-58.633	-17.267	1991	gbif
Cercomacra_melanaria	-57.571	-19.021	2019	inat
Cercomacra_melanaria	-58.5	-18.795	1969	UMSA- Gomez, I.
Cercomacra_melanaria	-58.191	-19.703	1971	UMSA- Gomez, I.
Cercomacra_melanaria	-58.102	-19.706	1972	UMSA- Gomez, I.
Cercomacra_melanaria	-58.116	-19.829	1973	UMSA- Gomez, I.
Cercomacra_melanaria	-58.221	-19.659	1977	UMSA- Gomez, I.
Cercomacra_melanaria	-57.992	-18.987	1979	UMSA- Gomez, I.
Cercomacra_melanaria	-57.905	-18.796	1982	UMSA- Gomez, I.
Cercomacra_melanaria	-57.962	-18.886	1986	UMSA- Gomez, I.
Cercomacra_melanaria	-58.554	-18.903	1989	UMSA- Gomez, I.
Cercomacra_melanaria	-58.572	-18.793	1990	UMSA- Gomez, I.
Cercomacra_melanaria	-57.804	-18.333	1991	UMSA- Gomez, I.
Cercomacra_melanaria	-57.858	-18.491	1992	UMSA- Gomez, I.
Cercomacra_melanaria	-58.798	-17.908	1992	UMSA- Gomez, I.
Cercomacra_melanaria	-58.872	-17.946	1993	UMSA- Gomez, I.
Cercomacra_melanaria	-58.667	-17.221	1993	UMSA- Gomez, I.
Cercomacra_melanaria	-58.771	-17.284	1993	UMSA- Gomez, I.
Cercomacra_melanaria	-60.317	-17.103	1994	UMSA- Gomez, I.
Cercomacra_melanaria	-60.267	-17.026	1995	UMSA- Gomez, I.
Cercomacra_melanaria	-60.181	-17.19	1995	UMSA- Gomez, I.
Cercomacra_melanaria	-62.169	-18.824	1996	UMSA- Gomez, I.
Cercomacra_melanaria	-61.998	-18.438	1996	UMSA- Gomez, I.
Cercomacra_melanaria	-61.684	-18.21	1997	UMSA- Gomez, I.
Cercomacra_melanaria	-62.312	-18.866	1997	UMSA- Gomez, I.
Cercomacra_melanaria	-62.883	-17.781	1998	UMSA- Gomez, I.
Cercomacra_melanaria	-62.655	-17.681	1999	UMSA- Gomez, I.
Cercomacra_melanaria	-62.698	-17.896	1999	UMSA- Gomez, I.
Cercomacra_melanaria	-63.383	-17.482	1999	UMSA- Gomez, I.
Cercomacra_melanaria	-63.411	-17.082	2000	UMSA- Gomez, I.
Cercomacra_melanaria	-63.559	-16.933	2000	UMSA- Gomez, I.
Cercomacra_melanaria	-64.862	-16.814	2000	UMSA- Gomez, I.
Cercomacra_melanaria	-64.892	-16.731	2000	UMSA- Gomez, I.
Cercomacra_melanaria	-64.785	-16.749	2001	UMSA- Gomez, I.

Cercomacra_melanaria	-64.505	-16.1	2004	UMSA- Gomez, I.
Cercomacra_melanaria	-64.517	-15.922	2005	UMSA- Gomez, I.
Cercomacra_melanaria	-64.594	-15.94	2005	UMSA- Gomez, I.
Cercomacra_melanaria	-64.035	-15.52	2005	UMSA- Gomez, I.
Cercomacra_melanaria	-64.041	-15.371	2005	UMSA- Gomez, I.
Cercomacra_melanaria	-64.13	-15.472	2005	UMSA- Gomez, I.
Cercomacra_melanaria	-64.404	-15.407	2006	UMSA- Gomez, I.
Cercomacra_melanaria	-64.344	-15.317	2006	UMSA- Gomez, I.
Cercomacra_melanaria	-64.469	-15.24	2006	UMSA- Gomez, I.
Cercomacra_melanaria	-64.755	-15.383	2007	UMSA- Gomez, I.
Cercomacra_melanaria	-64.999	-15.228	2007	UMSA- Gomez, I.
Cercomacra_melanaria	-64.927	-15.103	2007	UMSA- Gomez, I.
Cercomacra_melanaria	-65.564	-15.074	2008	UMSA- Gomez, I.
Cercomacra_melanaria	-64.541	-15.014	2008	UMSA- Gomez, I.
Cercomacra_melanaria	-64.678	-14.859	2008	UMSA- Gomez, I.
Cercomacra_melanaria	-64.428	-14.919	2008	UMSA- Gomez, I.
Cercomacra_melanaria	-64.582	-14.949	2008	UMSA- Gomez, I.
Cercomacra_melanaria	-64.672	-14.58	2009	UMSA- Gomez, I.
Cercomacra_melanaria	-64.112	-14.306	2009	UMSA- Gomez, I.
Cercomacra_melanaria	-63.482	-13.878	2009	UMSA- Gomez, I.
Cercomacra_melanaria	-63.797	-13.473	2009	UMSA- Gomez, I.
Cercomacra_melanaria	-63.857	-13.521	2009	UMSA- Gomez, I.
Cercomacra_melanaria	-64.285	-13.687	2010	UMSA- Gomez, I.
Cercomacra_melanaria	-64.374	-13.414	2010	UMSA- Gomez, I.
Cercomacra_melanaria	-64.428	-13.45	2010	UMSA- Gomez, I.
Cercomacra_melanaria	-64.749	-13.783	2011	UMSA- Gomez, I.
Cercomacra_melanaria	-64.921	-13.878	2011	UMSA- Gomez, I.
Cercomacra_melanaria	-64.856	-13.83	2012	UMSA- Gomez, I.
Cercomacra_melanaria	-64.648	-14.056	2012	UMSA- Gomez, I.
Cercomacra_melanaria	-64.808	-14.217	2012	UMSA- Gomez, I.
Cercomacra_melanaria	-64.761	-14.134	2012	UMSA- Gomez, I.
Cercomacra_melanaria	-64.606	-14.288	2012	UMSA- Gomez, I.
Cercomacra_melanaria	-64.678	-14.389	2012	UMSA- Gomez, I.
Cercomacra_melanaria	-65.623	-15.014	2013	UMSA- Gomez, I.
Cercomacra_melanaria	-65.951	-14.675	2013	UMSA- Gomez, I.
Cercomacra_melanaria	-66.052	-14.604	2014	UMSA- Gomez, I.
Cercomacra_melanaria	-66.307	-14.294	2014	UMSA- Gomez, I.
Cercomacra_melanaria	-66.409	-14.33	2014	UMSA- Gomez, I.
Cercomacra_melanaria	-66.432	-14.217	2014	UMSA- Gomez, I.
Cercomacra_melanaria	-66.313	-14.223	2014	UMSA- Gomez, I.
Cercomacra_melanaria	-66.926	-14.52	2014	UMSA- Gomez, I.

Cercomacra_melanaria	-66.908	-14.455	2015	UMSA- Gomez, I.
Cercomacra_melanaria	-67.027	-14.092	2015	UMSA- Gomez, I.
Cercomacra_melanaria	-67.009	-13.967	2015	UMSA- Gomez, I.
Cercomacra_melanaria	-65.885	-13.705	2015	UMSA- Gomez, I.
Cercomacra_melanaria	-67.072	-14.3	2015	UMSA- Gomez, I.
Cercomacra_melanaria	-67.015	-14.215	2016	UMSA- Gomez, I.
Cercomacra_melanaria	-67.101	-14.043	2016	UMSA- Gomez, I.
Cercomacra_melanaria	-66.951	-14.029	2016	UMSA- Gomez, I.
Cercomacra_melanaria	-66.815	-14.136	2016	UMSA- Gomez, I.
Cercomacra_melanaria	-66.587	-14.6	2016	UMSA- Gomez, I.
Cercomacra_melanaria	-66.48	-14.693	2016	UMSA- Gomez, I.
Cercomacra_melanaria	-66.194	-14.978	2017	UMSA- Gomez, I.
Cercomacra_melanaria	-66.094	-15.114	2017	UMSA- Gomez, I.
Cercomacra_melanaria	-66.244	-15.107	2017	UMSA- Gomez, I.
Cercomacra_melanaria	-65.067	-15.185	2017	UMSA- Gomez, I.
Cercomacra_melanaria	-65.145	-15.185	2017	UMSA- Gomez, I.
Cercomacra_melanaria	-64.524	-15.549	2018	UMSA- Gomez, I.
Cercomacra_melanaria	-64.824	-15.235	2018	UMSA- Gomez, I.
Cercomacra_melanaria	-65.067	-14.843	2018	UMSA- Gomez, I.
Cercomacra_melanaria	-65.052	-14.721	2018	UMSA- Gomez, I.
Cercomacra_melanaria	-65.052	-14.636	2018	UMSA- Gomez, I.
Cercomacra_melanaria	-64.981	-14.286	2018	UMSA- Gomez, I.
Cercomacra_melanaria	-65.074	-14.286	2018	UMSA- Gomez, I.
Cercomacra_melanaria	-65.352	-14.143	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-65.059	-13.95	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-65.002	-13.893	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-64.988	-13.986	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-65.466	-13.708	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-65.359	-13.622	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-65.445	-13.85	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-65.324	-13.779	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-64.631	-13.793	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-64.695	-13.722	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-64.595	-13.472	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-64.688	-13.565	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-64.703	-13.408	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-64.617	-13.272	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-64.681	-13.322	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-63.939	-13.251	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-63.946	-13.122	2017	UMSA - Lothersberger, S.
Cercomacra_melanaria	-63.882	-13.179	2018	UMSA - Lothersberger, S.

Cercomacra_melanaria	-63.767	-17.127	2018	UMSA - Lothersberger, S.
Cercomacra_melanaria	-63.71	-17.07	2018	UMSA - Lothersberger, S.
Cercomacra_melanaria	-63.782	-17.206	2018	UMSA - Lothersberger, S.
Cercomacra_melanaria	-63.896	-17.555	2015	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-63.953	-17.748	2015	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-64.024	-17.755	2015	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-63.989	-17.648	2015	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-63.189	-17.934	2016	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-63.175	-17.784	2016	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-63.268	-17.941	2016	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-62.689	-19.039	2013	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-62.618	-19.224	2013	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-62.618	-19.346	2013	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-62.668	-19.281	2013	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-62.561	-19.139	2013	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-62.653	-18.967	2013	UAGRM- Montenegro, M. A.
Cercomacra_melanaria	-61.875	-18.518	2016	ASBOR - Camacho, D.
Cercomacra_melanaria	-61.968	-18.518	2016	ASBOR - Camacho, D.
Cercomacra_melanaria	-61.875	-18.389	2016	ASBOR - Camacho, D.
Cercomacra_melanaria	-57.787	-19.068	2017	ASBOR - Camacho, D.
Cercomacra_melanaria	-57.77	-18.906	2017	ASBOR - Camacho, D.
Cercomacra_melanaria	-57.552	-18.247	2017	ASBOR - Camacho, D.
Cercomacra_melanaria	-58.592	-17.325	2017	ASBOR - Camacho, D.
Cercomacra_melanaria	-58.409	-17.191	2017	ASBOR - Camacho, D.
Cercomacra_melanaria	-58.663	-17.421	2017	ASBOR - Camacho, D.

Appendix 2

Table A2.2: Factorial analysis result with the 19 Bioclimatic variables.

Bioclimatic variables	MR1	MR4	MR2	MR3
BIO1 = Annual Mean Temperature	0.97	0.15	0.16	0.02
BIO2 = Mean Diurnal Range	0.68	0.47	0.06	0.24
BIO3 = Isothermality (BIO2/BIO7) (* 100)	0.14	0.18	0.19	0.88
BIO4 = Temperature Seasonality	0.04	0.1	0.27	0.96
BIO5 = Max Temperature of Warmest Month	0.95	0.02	0.14	0.2
BIO6 = Min Temperature of Coldest Month	0.92	0.26	0.17	0.24
BIO7 = Temperature Annual Range	0.47	0.45	0.15	0.67
BIO8 = Mean Temperature of Wettest Quarter	0.97	0.15	0.12	0.12
BIO9 = Mean Temperature of Driest Quarter	0.95	0.17	0.19	0.18
BIO10 = Mean Temperature of Warmest Quarter	0.98	0.14	0.12	0.11
BIO11 = Mean Temperature of Coldest Quarter	0.94	0.17	0.2	0.2
BIO12 = Annual Precipitation	0.26	0.45	0.79	0.32
BIO13 = Precipitation of Wettest Month	0.19	0.18	0.94	0.21
BIO14 = Precipitation of Driest Month	0.1	0.88	0.44	0.12
BIO15 = Precipitation Seasonality	0.44	0.8	0	0.18
BIO16 = Precipitation of Wettest Quarter	0.2	0.18	0.93	0.23
BIO17 = Precipitation of Driest Quarter	0.13	0.87	0.44	0.16
BIO18 = Precipitation of Warmest Quarter	0.17	0.44	0.75	0.02
BIO19 = Precipitation of Coldest Quarter	0.2	0.8	0.43	0.17

Appendix 3

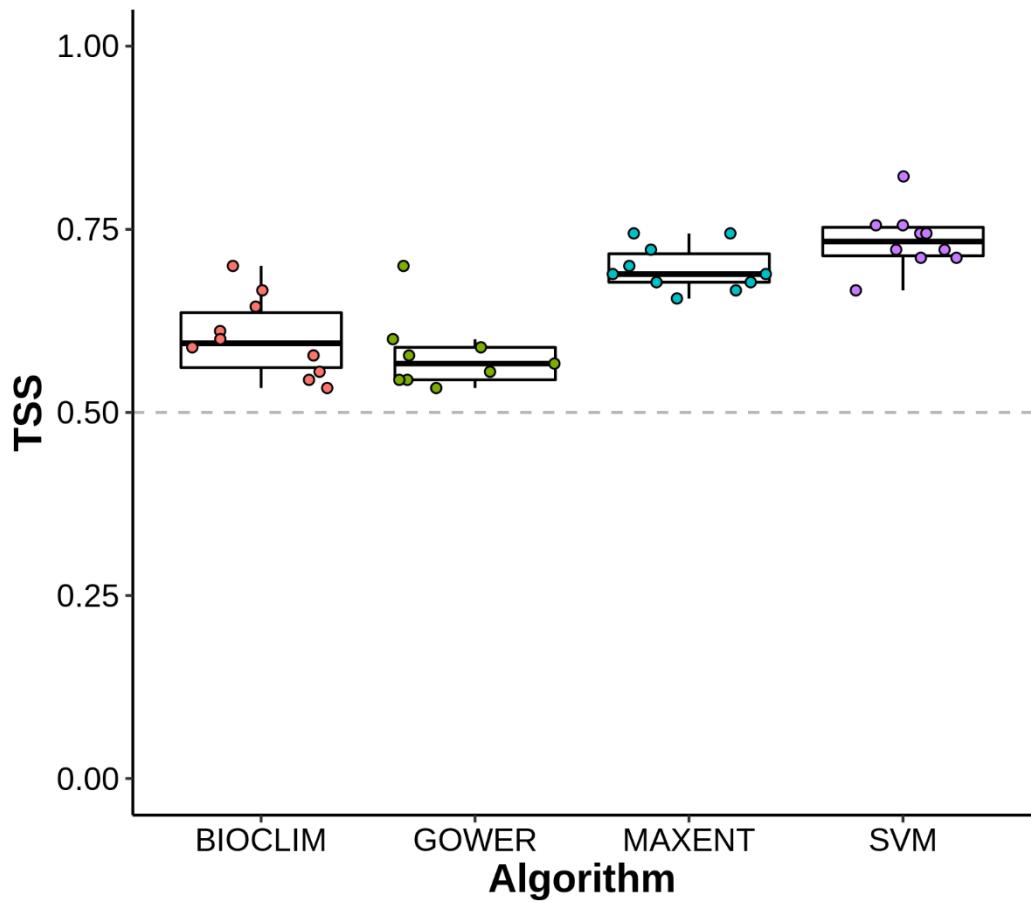
Table A3.3: Result of model evaluation according to TSS (true skill statistic), Threshold and AUC value for each algorithm.

	Algorithm	Threshold	AUC	TSS
	BIOCLIM	0.0475191	0.823	0.589
	BIOCLIM	0.0475191	0.7969	0.533
	BIOCLIM	0.0379952	0.8133	0.544
	BIOCLIM	0.0284714	0.8294	0.611
	BIOCLIM	0.0808524	0.8617	0.7
	BIOCLIM	0.0665667	0.813	0.644
	BIOCLIM	0.0713286	0.8563	0.6
	BIOCLIM	0.0237095	0.8898	0.667
	BIOCLIM	0.0475191	0.8101	0.556
	BIOCLIM	0.0284714	0.8252	0.578
	GOWER	0.5311748	0.8497	0.589
	GOWER	0.5288443	0.7622	0.556
	GOWER	0.5289337	0.7698	0.544
	GOWER	0.5206949	0.8111	0.567
	GOWER	0.551321	0.8322	0.6
	GOWER	0.5283914	0.8126	0.578
	GOWER	0.5099529	0.7926	0.544
	GOWER	0.5359615	0.8822	0.7
	GOWER	0.5234998	0.8028	0.533
	MAXENT	0.429548	0.8737	0.689
	MAXENT	0.5373038	0.8863	0.722
	MAXENT	0.4750426	0.869	0.667
	MAXENT	0.4994759	0.8674	0.678
	MAXENT	0.471922	0.8673	0.744
	MAXENT	0.4417572	0.8727	0.7
	MAXENT	0.4527299	0.8936	0.678
	MAXENT	0.498503	0.9109	0.744
	MAXENT	0.5093525	0.8811	0.689
	MAXENT	0.4060642	0.8758	0.656
	SVM	0.618196	0.9141	0.722
	SVM	0.6344363	0.9065	0.744
	SVM	0.6196699	0.8826	0.711
	SVM	0.707109	0.9181	0.744
	SVM	0.4946061	0.8958	0.756
	SVM	0.4008957	0.8931	0.722
	SVM	0.6751981	0.9131	0.756

SVM	0.712212	0.9372	0.822
SVM	0.8114853	0.8868	0.667
SVM	0.6167689	0.8836	0.711

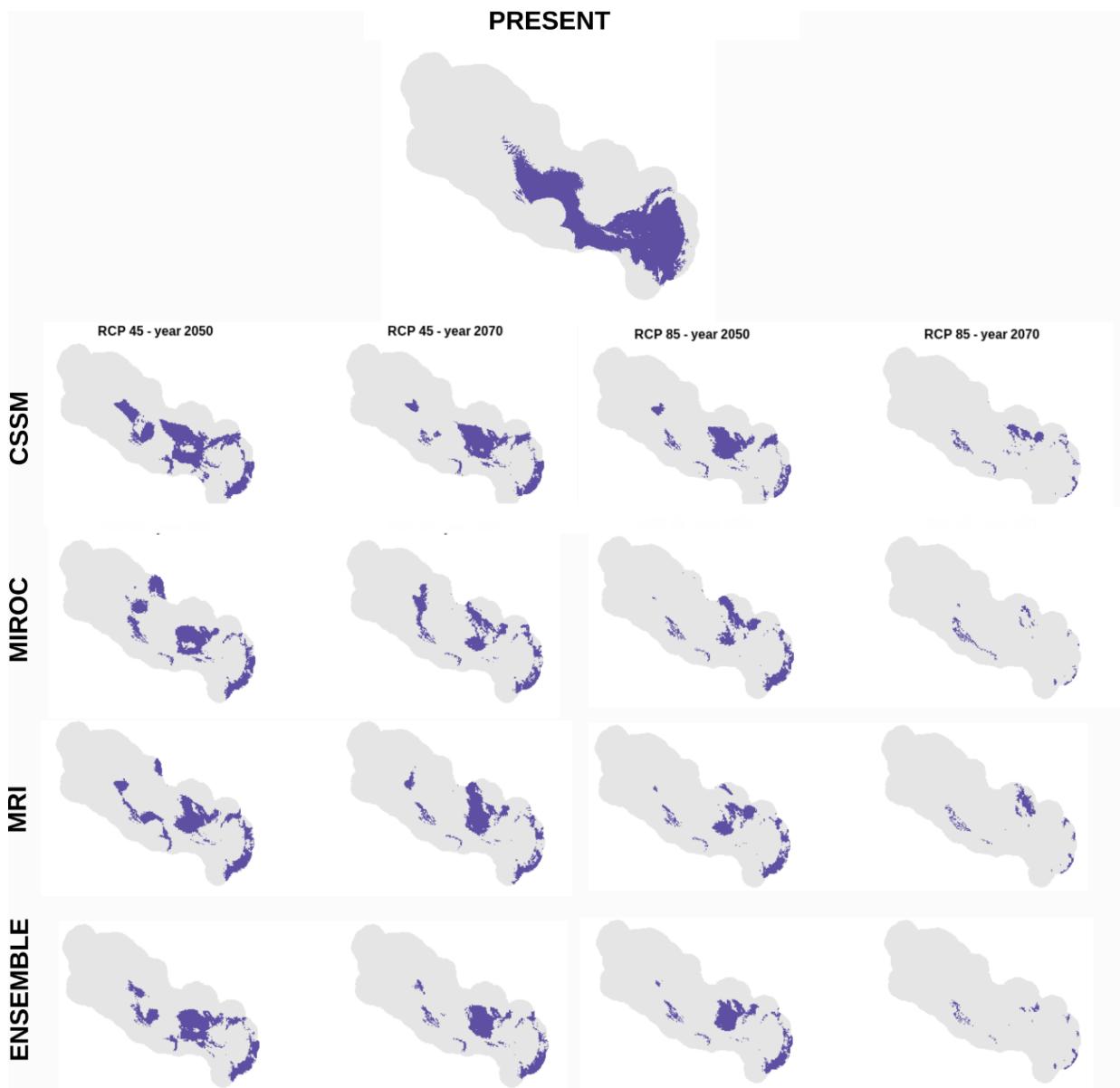
Appendix 4

Fig.A4.4: Result of model evaluation according to TSS (true skill statistic), Threshold and AUC value for each algorithm.



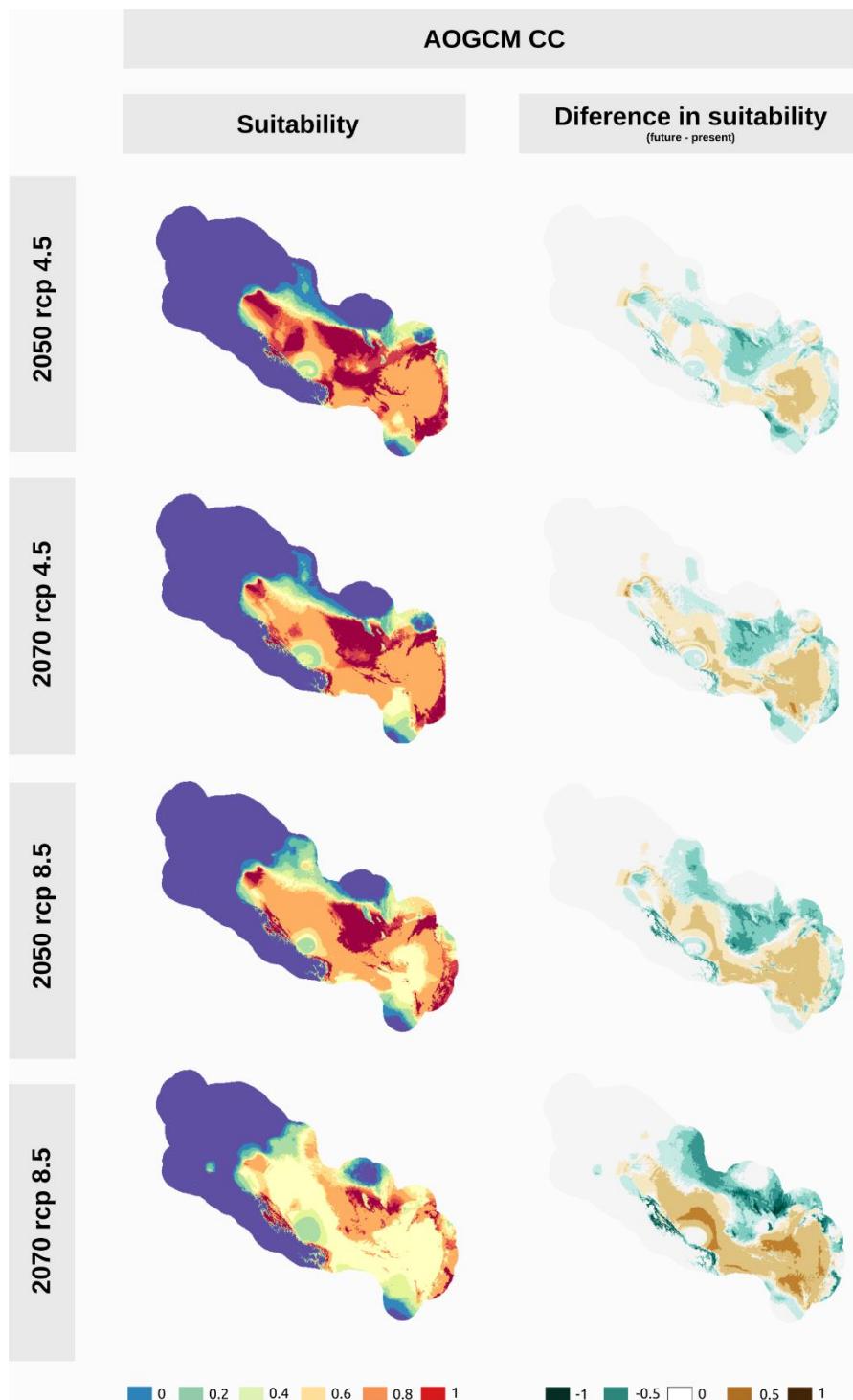
Appendix 5

Fig.A5.5: Potential distribution map of Mato Grosso Antbird suitability in different climate scenarios (present, year 2050, year 2070), different global circulation models (GCM) (CCSSM4, MIROC and MRI) and ensemble of the three GCMs.



Appendix 6

Fig.A6.6: Potential distribution maps of Mato Grosso Antbird suitability in different climate scenarios (year 2050, year 2070) for each different global circulation models CCSM4, MIROC and MRI. (RCP: Representative Concentration Pathway).

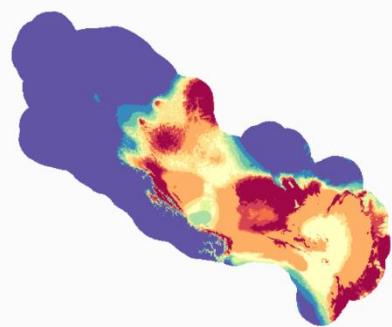


AOGCM MIROC

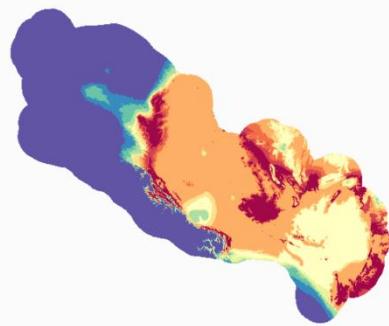
Suitability

Difference in suitability
(future - present)

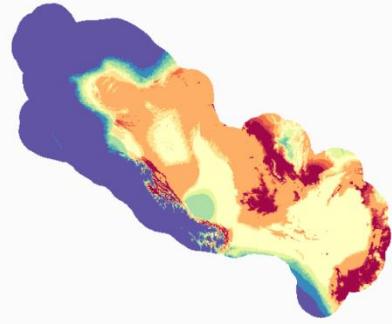
2050 rcp 4.5



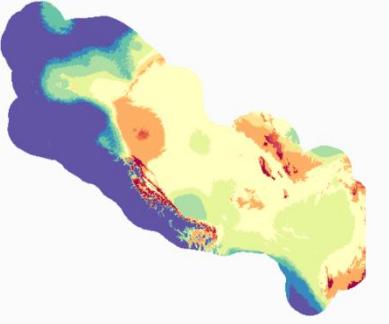
2070 rcp 4.5



2050 rcp 8.5

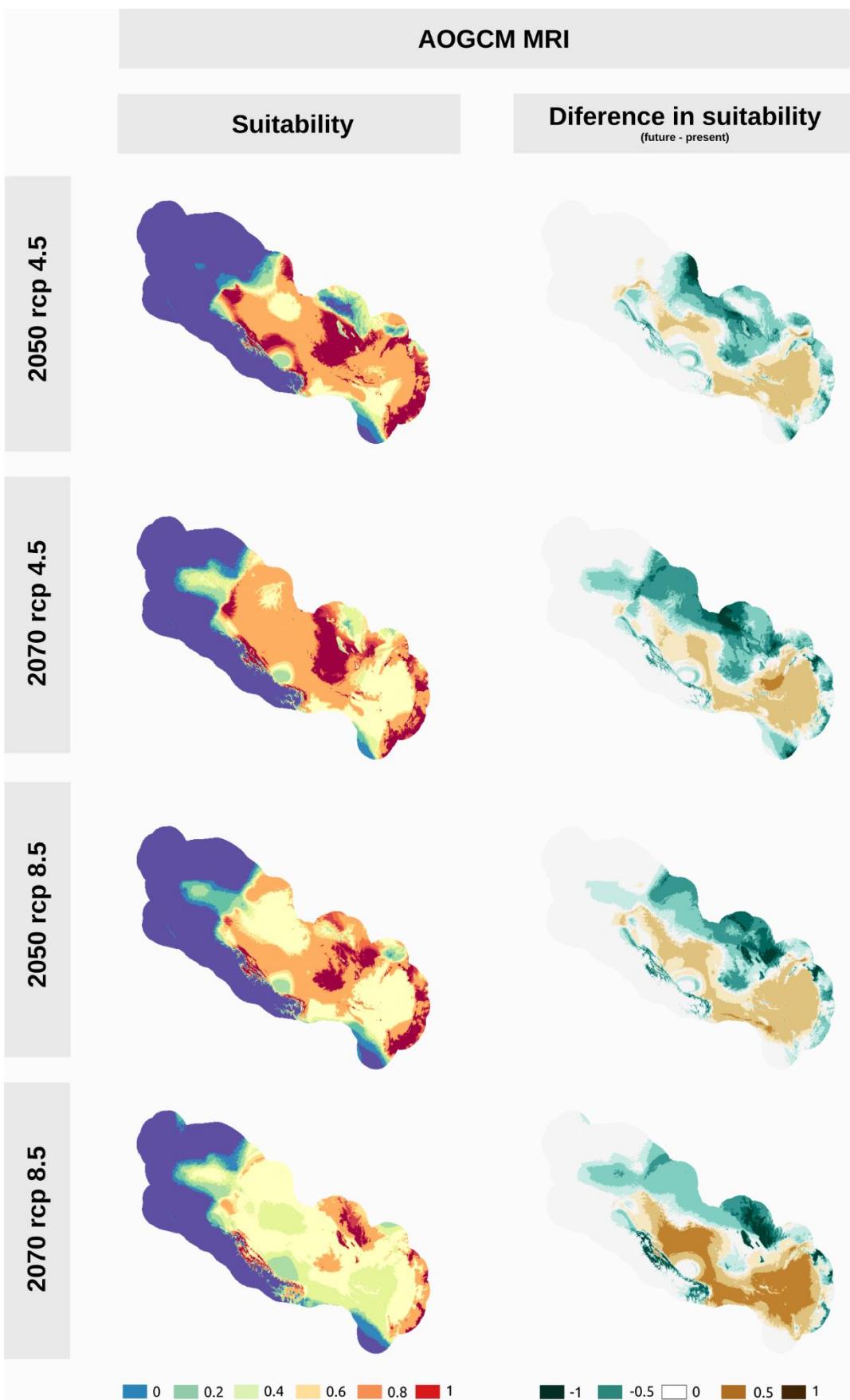


2070 rcp 8.5



0 0.2 0.4 0.6 0.8 1

-1 -0.5 0 0.5 1



Appendix 7

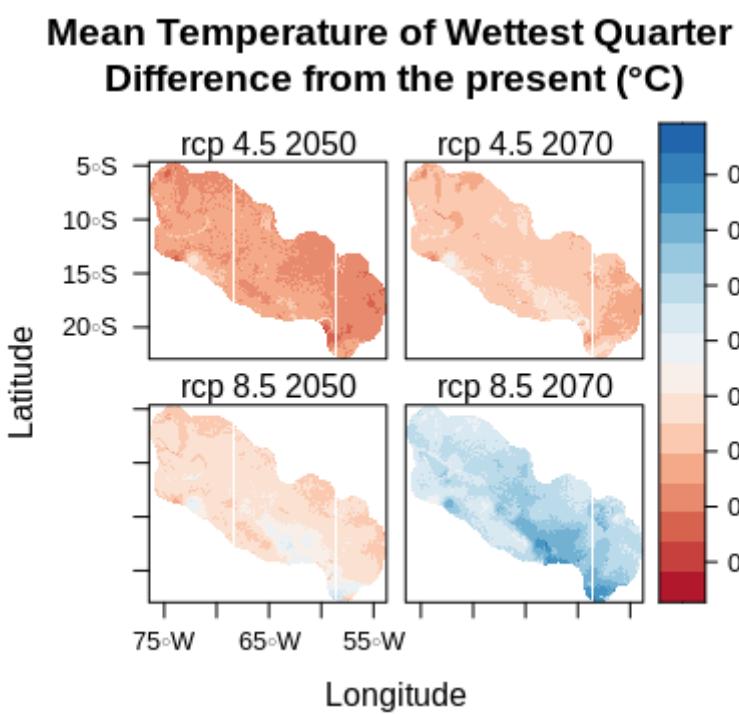
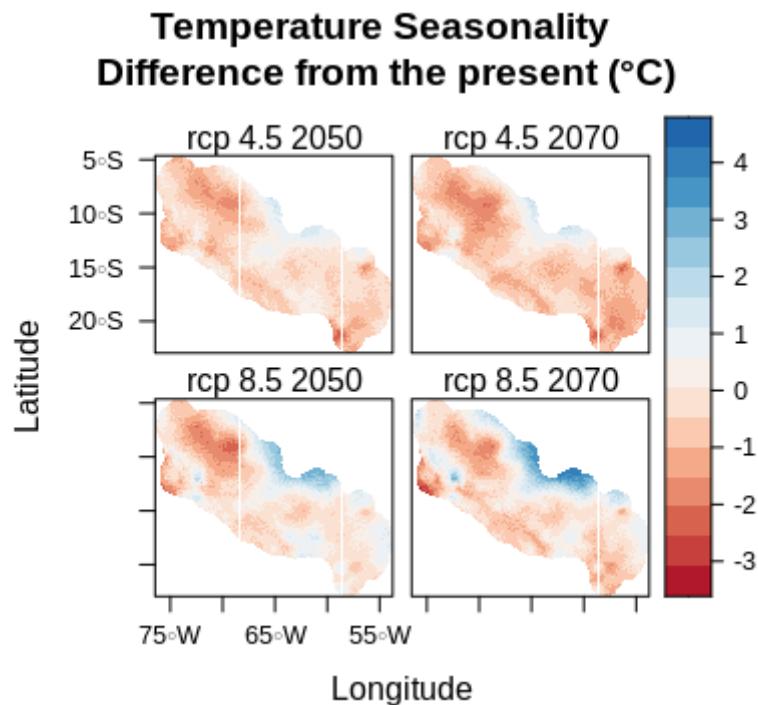
Table A7.7: Results of comparison of different climate scenarios for each dependent variable (elevation, slope, latitude and longitude) performed by post-hoc Dunn test. Variable - Variables compared in post-hot Dunn test (elevation, slope, latitude and longitude); Samples - Present day and future samples (2050 and 2070) compared in the two future contrasting emission scenarios of the Intergovernmental Panel on Climate Change (IPCC), Fifth Assessment Report-Coupled Model Intercomparison Project, Phase 5 (AR-CMIP5) / Representative Concentration Pathway (RCP) 4.5 - lower energy demand, and IPCC AR-CMIP 5 / RCP 8.5 - maximum power requirement, balanced emissions from fossil fuels and non-fossil fuels; post-hoc Dunn test statistic; *P*-adjusted.

Variable	Comparison	Z	P.adjusted
Elevation	Present - 2050 rcp4.5	-97.897	<0.001
Elevation	Present - 2070 rcp4.5	-111.98	<0.001
Elevation	2050 rcp 4.5 - 2070 rcp 4.5	-22.046	<0.001
Elevation	Present - 2050 rcp 8.5	-111.702	<0.001
Elevation	2050 rcp 4.5 - 2050 rcp 8.5	-25.583	<0.001
Elevation	2070 rcp 4.5 - 2050 rcp 8.5	-4.08	<0.001
Elevation	Present - 2070 rcp 8.5	-93.668	<0.001
Elevation	2050 rcp 4.5 - 2070 rcp 8.5	-52.084	<0.001
Elevation	2070 rcp 4.5 - 2070 rcp 8.5	-40.426	<0.001
Elevation	2050 rcp 8.5 - 2070 rcp 8.5	-37.871	<0.001
Slope	Present - 2050 rcp 4.5	-84.449	<0.001
Slope	Present - 2070 rcp 4.5	-93.561	<0.001
Slope	2050 rcp 4.5- 2070 rcp 4.5	-16.414	<0.001
Slope	Present - 2050 rcp 8.5	-93.6	<0.001
Slope	2050 rcp 4.5 - 2050 rcp 8.5	-19.669	<0.001
Slope	2070 rcp 4.5 - 2050 rcp8.5	-3.607	0.003
Slope	Present - 2070 rcp 8.5	-65.895	<0.001
Slope	2050 rcp 4.5 - 2070 rcp 8.5	-31.141	<0.001
Slope	2070 rcp 4.5 - 2070 rcp 8.5	-22.768	<0.001
Slope	2050 rcp 8.5 - 2070 rcp 8.5	-20.674	<0.001
Latitude	Present - 2050 rcp 4.5	-22.443	<0.001
Latitude	Present - 2070 rcp 4.5	-23.132	<0.001
Latitude	2050 rcp 4.5 - 2070 rcp 4.5	-2.725	0.064
Latitude	Present - 2050 rcp 8.5	-20.711	<0.001
Latitude	2050 rcp 4.5 - 2050 rcp 8.5	-1.343	1
Latitude	2070 rcp 4.5 - 2050 rcp 8.5	1.192	1
Latitude	Present - 2070 rcp 8.5	-10.259	<0.001
Latitude	2050 rcp 4.5 - 2070 rcp 8.5	-0.944	1
Latitude	2070 rcp 4.5 - 2070 rcp 8.5	0.391	1
Latitude	2050 rcp 8.5 - 2070 rcp 8.5	-0.249	1

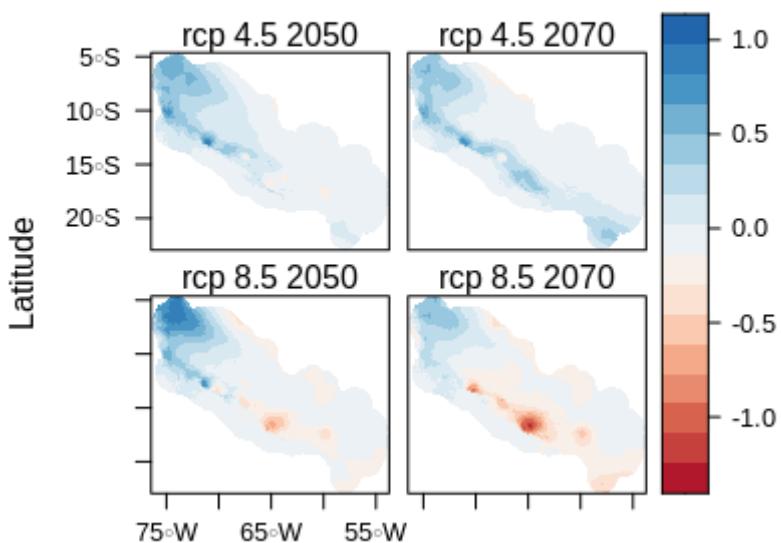
Longitude	Present - 2050 rcp 4.5	4.122	<0.001
Longitude	Present - 2070 rcp 4.5	-8.497	<0.001
Longitude	2050 rcp 4.5 - 2070 rcp 4.5	-11.188	<0.001
Longitude	Present - 2050 rcp 8.5	-18.589	<0.001
Longitude	2050 rcp 4.5 - 2050 rcp 8.5	-20.434	<0.001
Longitude	2070 rcp 4.5 - 2050 rcp 8.5	-9.018	<0.001
Longitude	Present - 2070 rcp 8.5	-5.784	<0.001
Longitude	2050 rcp 4.5 - 2070 rcp 8.5	-7.345	<0.001
Longitude	2070 rcp 4.5- 2070 rcp 8.5	-1.795	0.727
Longitude	2050 rcp 8.5 - 2070 rcp 8.5	3.038	0.024

Appendix 8

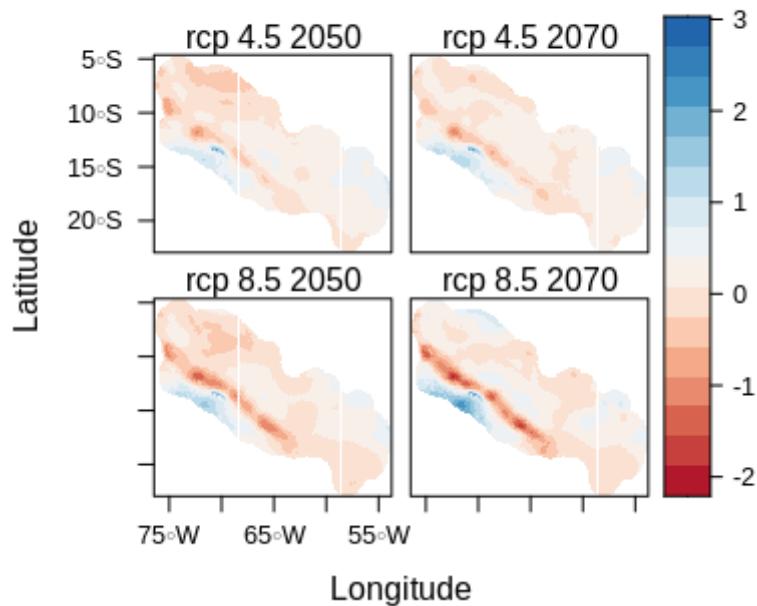
Fig.A8.8: Difference between current and future of the four bioclimatic variables used in niche modeling. These values were extracted from the raster used in this study. (RCP: Representative Concentration Pathway)



Precipitation of Driest Quarter Difference from the present (mm)



Precipitation of Wettest Quarter Difference from the present (mm)

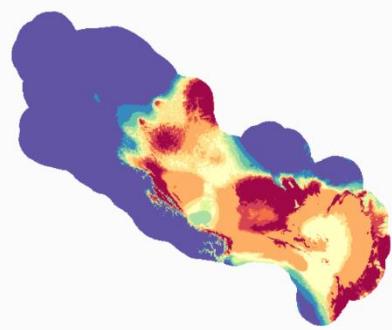


AOGCM MIROC

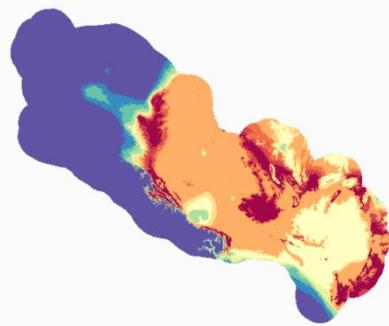
Suitability

Difference in suitability
(future - present)

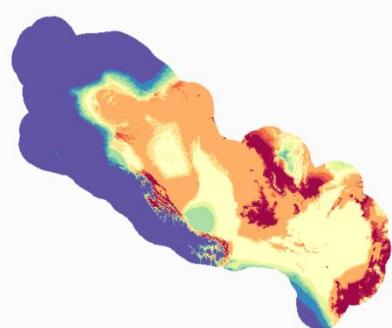
2050 rcp 4.5



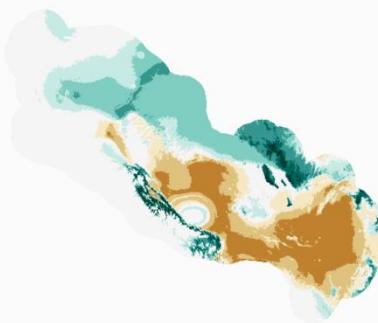
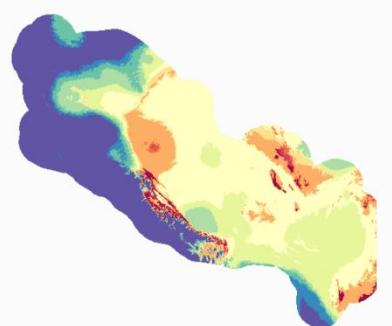
2070 rcp 4.5



2050 rcp 8.5



2070 rcp 8.5



0 0.2 0.4 0.6 0.8 1

-1 -0.5 0 0.5 1

