# Potential future climate change effects on global reptile distribution and diversity

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- 7 **Running head:** Global reptiles under climate change
- 8 Abstract

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- 9 **Aim:** Until recently, complete information on global reptile distributions has not been widely
- 10 available. Here, we provide the first comprehensive climate impact assessment for reptile
- 11 distributions at a global scale.
- 12 **Location:** Global, excluding Antarctica
- **13 Time period:** 1995, 2050, 2080
- 14 **Major taxa studied:** Reptiles
- 15 **Methods:** We performed species distribution models for 6296 reptile species and assessed
- potential global as well as realm-specific changes in species richness, the change in global
- 17 species richness across climatic space and species-specific changes in distribution and range
- 18 extent and overlap, under future climate change. To assess the future climatic impact of 3768
- 19 non-modeled species, we compared the future change in climatic conditions between both
- 20 modeled and non-modeled species.
- 21 **Results:** Reptile richness was projected to decline significantly over time, globally but also for
- 22 most zoogeographic realms, with the strongest decrease in Brazil, Australia and South Africa.
- 23 Species richness was highest in warm, but moist regions, which were projected to shift further to
- 24 climate extremes in the future. Extents of occurrence were projected to decline considerably in
- 25 the future, with a low overlap between projected current and future ranges. Shifts in range
- 26 centroids differed among realms and taxa, with a dominating global poleward shift. Non-
- 27 modeled species were significantly stronger affected by climatic changes than modeled species.
- **Main conclusions:** Reptile richness was projected to decrease significantly across most parts of
- 29 the world with ongoing future climate change. This effect is visible across lizards, snakes and
- 30 turtles alike and has considerable impact on species' extent of occurrence (EOO) and range
- 31 distribution. Together with other anthropogenic impacts, such as habitat loss and harvesting, this

- 32 is cause for concern. Given the historical lack of information on global reptile distributions, this
- 33 calls for an re-assessment of global conservation efforts towards reptile species, with specific
- 34 focus on anticipated future climatic changes.
- 35 **KEYWORDS:** species distribution model, turtle, snake, lizard, ISIMIP, bioclim, global
- 36 warming, environmental niche model

# Introduction

- 38 Emissions from anthropogenic activities have lead to an increase in global surface temperature of
- 39 around 1°C in the last 100 years. This has already led to changes in weather and climate
- 40 extremes in every region across the globe (IPCC, 2021). Unless emissions are vastly reduced in
- 41 the coming decades, global warming will continue and exceed 1.5°-2°C compared to pre-
- 42 industrial levels by the end of the 21st century (IPCC, 2021).
- 43 Climate change has already had adverse effects on biodiversity and ecosystem functioning and
- 44 these effects are likely to worsen as warming proceeds in future (IPBES, 2019; IPCC, 2022).
- 45 Climate change impacts on ecological processes scales from genes to entire ecosystems, can
- 46 affect organisms, populations or entire communities and vary between physiological,
- 47 morphological, phenological and distributional shifts (Bellard *et al.*, 2012; Scheffers *et al.*,
- 48 2016). Especially changes in species abundance and distribution due to climate change have
- 49 already been frequently observed (Bowler *et al.*, 2017; Lenoir *et al.*, 2020), with many species
- 50 shifting their range towards higher latitudes and elevations (Chen et al., 2011). However, some
- 51 species also respond to climate change by idiosyncratic range shifts (Gibson-Reinemer & Rahel,
- 52 2015).
- 53 Species distribution models (SDMs) are a common way of assessing species-specific responses
- 54 to climate change (e.g. Engelhardt *et al.* (2020)), but also to assess climate change impacts on
- 55 biodiversity (Thuiller *et al.*, 2005). SDMs statistically infer a relationship between the observed
- distribution of a species to the underlying climatic conditions (Elith & Leathwick, 2009) and can
- 57 then be used to project current distributions into the future (Elith et al., 2010), assuming that the
- 58 species maintains its climatic niche (Wiens & Graham, 2005). By doing this for multiple species,
- 59 these projections can be combined to assess future changes in species richness (Hof *et al.*, 2018;
- 60 Newbold, 2018; Thuiller *et al.*, 2019).
- In the past most climate change impact assessments on vertebrate biodiversity have focused on
- 62 endotherms (birds & mammals). Reptiles, though accounting for a third of global terrestrial
- 63 vertebrate diversity, have been largely ignored (Pacifici et al., 2015). Previous studies that
- assessed climate change impacts on reptiles species, have either only used a subset of species
- 65 (Warren et al., 2018; Newbold, 2018) or have not been of global extent (Araújo et al., 2006).
- 66 Until recently, this was largely due to the unavailability of global reptile distribution data, but
- 67 this has changed with the release of the global distribution database by the Global Assessment of
- Reptile Distributions (GARD) initiative (Roll et al., 2017) and the release of the full set of IUCN
- 69 reptile range maps (IUCN, 2022).
- 70 Global biodiversity assessments further often either consider overall effects on a single taxon
- 71 (Baisero et al., 2020; Voskamp et al., 2021) or compare multiple taxa (Hof et al., 2018;
- Newbold, 2018; Warren et al., 2018; Thuiller et al., 2019), but only very rarely compare
- 73 different taxonomic groups within one taxon (but see, for example Hof et al. (2011)). This may

- be problematic as pooling all species may obscure the evolutionary and biogeographic history of
- 75 major lineages, and from a global conservation perspective ignores the fact that hotspots of total
- 76 reptile richness hardly overlap with those of lizard or turtle richness (Roll *et al.*, 2017).
- Here, we tried to fill the gaps outlined above, by providing a detailed account of projected
- 78 climate change impacts on global reptile distributions and diversity, looking at species-specific
- 79 changes as well as broad-scale geographic trends across and within different taxonomic groups.
- 80 We assessed changes in reptile richness globally, within each zoogeographic realm and across
- 81 their respective climate space. For each species, we further quantified the change in extent of
- 82 occurrence (EOO), range overlap and range distribution and again assessed differences across
- 83 zoogeographic realms and taxonomic groups. Given that we cannot model range restricted
- 84 species, we also performed a more general assessment of species-specific changes in climatic
- space across both modeled and non-modeled species.

# Methods

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## Species data

- We obtained global range maps of 10,064 reptile species from the Global Assessment of Reptile
- 89 Distributions (GARD, https://doi.org/10.5061/dryad.83s7k). Roll *et al.* (2017) provide a detailed
- 90 description of the methodology used for deriving the range maps. The range maps cover lizards,
- 91 snakes, turtles, worm lizards, crocodiles and the tuatara, but in this paper, similar to (Roll et al.,
- 92 2017), we only contrast snakes, turtles and paraphyletic lizards (the latter of which we
- 93 subsequently refer to as lizards for simplicity).
- Page Range maps were gridded to a 0.5° x 0.5° grid in WGS84, considering any cell that intersected
- 95 with the range polygon, and pseudo-absence data for each species were generated by randomly
- 96 selecting absences using a distance-weighted approach (see Hof *et al.*, 2018). The number of
- absences was either equal to the number of presences or 1000 absences for species with less than
- 98 1000 presences. For each species we derived 10 sets of pseudo-absences, which were modeled
- 99 separately.

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#### Climate data

- 101 Global bias-corrected daily climate (minimum temperature, maximum temperature and
- precipitation) data at a spatial resolution of 0.5° (WGS84) was obtained from the meteorological
- 103 forcing dataset 'EartH2Observe, WFDEI and ERA-Interim data Merged and Bias-corrected for
- 104 ISIMIP' (EWEMBI; Lange, 2016) for current conditions (1980 2009) and from the Inter-
- Sectoral Impact Model Intercomparison Project phase 2b (ISIMP2b; Frieler et al., 2017) for
- 106 future simulations (2036 2065 & 2066 2095). Future climate simulations were available from
- 107 four global circulation models (GCMs; GFDL-ESM2M, HadGEM2-ES, IPSL-CM5A-LR and
- 108 MIROC5) and for three representative concentration pathways (RCPs; RCP2.6, RCP6.0 and
- 109 RCP8.5). Monthly means of each climatic variable over the respective 30-year time periods,
- centered around 1995, 2050 and 2080, and for each future scenario (GCM & RCP) were used to
- calculate 19 bioclimatic variables using the 'dismo' package (Hijmans et al., 2021) in R (R Core
- 112 Team, 2021). We used the same model selection approach as described in Hof *et al.* (2018) and
- 113 fitted our models using the best-performing combination of four explanatory variables, which
- was temperature seasonality, maximum temperature of the warmest month, annual precipitation
- and precipitation seasonality.

# **Species Distribution Models (SDMs)**

- 117 Projections based on SDMs vary considerably among the model algorithm considered, for this
- we fitted two modeling algorithms with a good performance and discrimination capacity
- 119 (Meynard & Quinn, 2007; Elith et al., 2010), an additive (Generalized Additive Model (GAM))
- and a regression tree based model (Generalized Boosted Regression Models (GBM)).
- 121 GAMs were fitted with a Bernoulli response, a logit link and thin-plate regression splines using
- the 'mgcv' package (Wood, 2003, 2011) in R (R Core Team, 2021). GBMs were fitted with the
- 123 'gbm' package (Greenwell et al., 2020) in R (R Core Team, 2021) and the optimal parameter
- settings for learning rate (0.01 and 0.001), tree complexity (1, 2 and 3) and number of trees
- 125 (1000-10000) for each species were identified by cross-validation (Bagchi *et al.*, 2013).
- 126 Spatial autocorrelation in species distributions can bias parameter estimates and error
- probabilities (Kühn, 2007). Two different methods were used to account for spatial
- autocorrelation in the SDMs. Species with equal or more than 50 presences were modeled using
- an ecoregion-blocking approach. Here the world was divided into 10 blocks, based on a
- representative subset of the climatic space of each of the world's ecoregions (Olson *et al.*, 2001),
- and 10 models were built leaving out one block at a time, using the left out block for model
- evaluation (Bagchi et al., 2013). For range-restricted species (10 49 presences), we split the
- data into 10 datasets by repeatedly randomly selecting 70% of the data, using the left-out 30%
- for model evaluation. Species occurring in less than 10 grid cells were not modeled (N = 3602,
- 135 Table 1).

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- 136 The performance of the fitted SDMs was evaluated by calculating the overall AUC for each
- species (the average AUC across the 10 blocks and the 10 sets of pseudo-absences). Models with
- an overall AUC smaller than 0.7 were dropped (N = 166), which left us with SDMs for 6296
- reptile species (see Fig. S1.1 in Appendix S1 in Supporting Information), which represents 62.6
- 140 % of the total number of available species by GARD (Table 1).
- 141 The same modeling approach has been used previously to assess climate change impacts on
- birds, amphibians and reptiles, see Hof et al. (2018) and Biber et al. (2020). The former provides
- a more detailed explanation of the modeling methodology, while the latter gives a thorough
- account of the caveats and uncertainties associated with species distribution models.

#### 145 Future projections

- 146 Future species distributions were derived by predicting the models using the future bioclimatic
- variables of the two future time periods (2050, 2080) and the respective future scenario (GCM &
- RCP). Future projections of each species were limited to their original and the extent of their
- neighboring ecoregions, to avoid predictions to areas with analogue climatic conditions. Future
- projections were further limited by applying a species-specific dispersal buffer. For most species
- 151 considered here species-specific dispersal distances are still unknown (Nathan et al., 2012),
- hence we used species-specific dispersal buffers, which were based on the diameter (d) of the
- largest range polygon of a species. We used three species-specific dispersal scenarios (d/4, d/8,
- d/16, see Fig. S1.2 in Appendix S1) and provide a detailed comparison of these in the Supporting
- 155 Information (see Appendix S4), but below provide the results under the medium dispersal
- scenario d/8, with a mean dispersal distance of 2.4 km per year.

#### Impact analysis

- 158 The current and future probabilities of occurrence of the individual SDMs were thresholded into
- 159 binary presence-absence data using species-specific thresholds according to the true skill statistic
- 160 (MaxTSS; Allouche et al., 2006). Thresholded species occurrences were then used to calculate
- 161 current and future species richness, as well as richness increase, decrease, change and relative
- 162 change (%). Richness increase and decrease was identified by using the presence information of
- each individual species and then summing up the number of species that newly occur in a given
- grid cell (species increase) or species that disappear from the respective grid cell (species
- 165 decreae).

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- As stacked thresholded data frequently overestimates species richness, we also present the results
- using the stacked raw probabilities of occurrence, as suggested by Calabrese *et al.* (2014), of the
- individual SDMs without thresholding where applicable, in the Supporting Information (see
- 169 Appendix S3).
- 170 We calculated the overall mean projected species richness globally and for each zoogeographic
- realm, as defined by Holt *et al.* (2013), for each time period and tested for significant changes in
- 172 species richness over time using a paired t-test with Holm correction. To assess how species
- 173 richness and changes in species richness are related to the overall change in climatic conditions,
- we assessed both across univariate temperature and precipitation as well as the interaction of
- temperature and precipitation conditions. To assess potential future climate effects on individual
- species, we quantified the proportion of change in EOO and the proportion of range overlap for
- species. And to assess the geographic change between each current and future species range, we
- 178 calculated the range centroid for current and future conditions of each species and then identified
- the distance and direction of change between current and future range centroids.
- 180 Given that 37.4% of reptile species for which data was available could not be modeled (largely
- due to their restricted range extent, Table 1), we performed an additional analysis considering all
- 182 10,064 species for which data were available. We used the same 4 bioclimatic variables used for
- the SDMs to transform the multidimensional climate data in a two-dimensional climate space
- using the first two axes of a principal component analysis (PCA). PCAs were performed for both
- current and future conditions, considering the same GCMs, RCPs and time periods as before
- 186 (Fig. S1.3). For each scenario combination, we then calculated the Euclidean distance between
- the two PCA-axes of current and future conditions, to get a measure of climatic change (Fig.
- 188 S1.4). We then extracted the climatic distance for the gridded locations of each species and
- 189 compared the climatic distance of modeled and non-modeled species using a non-paired t-test
- 190 with Holm correction.
- Where no specific groups (lizards, snakes, turtles) are mentioned, we present the results for all
- 192 reptile species together. Results are presented as the ensemble mean, across the four GCMs and
- two model algorithms considered, for the year 2080 under a medium representative
- 194 concentration pathway (RCP6.0). A sensitivity analysis with regards to the variation across years
- and RCPs is shown in the Supporting Information (see Appendix S4).

#### Results

- 197 Projected reptile richness for current conditions varied between 0 (high latitudes) and 251 in the
- tropics, with particular hotspots in Brazil, Cameroon and Indonesia (Fig. 1a). Overall, reptile

- richness was dominated by lizard species (N = 3695), followed by snakes (N = 2305), while
- 200 turtle species only contributed marginally to the total number of modeled species (N = 296,
- Table 1). Looking at the spatial configuration of species richness across the three groups, snakes
- had the highest mean species richness ( $\mu_{mean} = 30.4 \pm 0.15$ ), followed by lizard ( $\mu_{mean} = 24.3 \pm 0.15$ )
- 203 0.09 SE) and turtle richness ( $\mu_{mean} = 3.71 \pm 0.02$  SE), while the mean total richness was 58.4  $\pm$
- 204 0.24 (SE; Fig. 1b, Fig. S2.5, Appendix S2). A large number of reptile species was projected to
- 205 disappear and at the same time a large number of new species were projected to appear in Brazil
- and Australia, while other regions showed either a strong species decrease or increase (Fig. 1
- 207 c,d). The strongest future species decreases were projected east of the Caspian Sea and in South
- 208 Africa (Fig. 1 c), while strong future increases were predicted in the south-west of China and in
- 209 the eastern United States (Fig. 1 d). Overall, species decrease was stronger than species increase,
- 210 which resulted in a stronger negative net change in species richness from 1995 to 2080 (Fig.
- 211 1c,d,e). The lowest negative net change in species richness was observed in Brazil, Australia and
- 212 South Africa, while the highest positive net change was projected for south-west China and the
- 213 western United States (Fig. 1 e). Relative change (%) was projected to be negative in particular
- 214 for most of the southern hemisphere, while the high northern latitudes showed a strong positive
- 215 relative change (Fig. 1 f).
- 216 Spatial patterns in species richness changes varied strongly across the three taxa, with lizards
- seeing both strong increases and decreases in Australia, snakes showing a strong decrease in
- 218 South America and turtles seeing a strong increase in eastern North America (Fig. S2.6). All
- 219 three taxa showed a positive net change in species richness in northern latitudes (Fig. S2.7),
- while lizards decreased in parts of Australia (Fig. S2.7 a,b), snakes in large parts of South
- America (Fig. S2.7 c, d) and turtles showed a strong negative net change in parts of South
- 222 America and southern Africa (Fig. S2.7 f).
- Globally reptile richness was projected to decline significantly (p < 0.01) from 1995 to 2080,
- with a decline in mean reptile richness from  $58.4 \pm 0.24$  (SE) in 1995 to  $53.39 \pm 0.19$  (SE) in
- 225 2080 (Fig. 2 a). 8 out of 11 zoogeographic realms showed a significant decline in reptile richness
- by 2080 (Fig. 2 b, c, d, f, g, h, j, k), while the Nearctic and Palearctic realm showed a significant
- increase (Fig. 2 e, i) and the Sino-Japanese realm showed no significant change (Fig. 2 l).
- 228 The different taxonomic groups (lizards, snakes and turtles) showed changes in global species
- 229 richness that were similar to all reptiles, while there were slight differences across the individual
- 230 realms. Lizards only showed a significant increase in richness in the Palearctic realm and no
- 231 significant change in richness in the Sino-Japanese realm, while in all other realms they showed
- a significant decrease (Fig. S2.8). Snake and turtle richness increased significantly in the
- 233 Nearctic and Palearctic realm. Snake richness decreased significantly in all other realms apart
- from the Sino-Japanese one, while turtle richness significantly decreased in all other realms apart
- from the Saharo-Arabian and the Sino-Japanese one (Fig. S2.9, S2.10).
- 236 Reptile richness strongly differed across conditions of varying combinations of temperature and
- precipitation (Fig. 3 a,b,c). For 1995 reptile richness was projected to be highest in areas with a
- 238 temperature around 28.5°C, a precipitation of about 5500 mm and when considering temperature
- and precipitation together in warm, but moist regions (21°C & 3000 mm, Fig. 3 c). The climatic
- 240 conditions with the highest richness shifted to even more extreme (warmer & wetter) novel
- 241 climate conditions by 2080 (Fig. 3 a,b,d).

- 242 Looking at the species richness change across the 2-dimensional climate space, net change was
- 243 positive at the upper precipitation limits across all temperatures and the very hot and very dry
- 244 conditions and negative throughout the entire precipitation range especially for the higher
- temperatures. Overall the negative change was much stronger and more pronounced than the
- positive net change (Fig. 3 e). The highest positive and negative relative change values were
- 247 clustered, both occurred at the upper precipitation limits at low and medium temperatures (Fig. 3
- 248 f). A considerable proportion of the climate space (29.5 %) was shifting towards novel climatic
- 249 conditions, for which no change in species richness could be estimated, while only few discrete
- 250 climatic conditions as well as very cold & very dry conditions (4.75 %) got lost (Fig. 3 e,f).
- 251 The EOO of most species (n = 6021) showed a considerable decrease ( $\mu$ mean = -27.7  $\pm$  0.16 SE,
- Fig. 4 a,c,e). Lizard species showed the strongest decline ( $\mu$ mean = -31.8  $\pm$  0.22 SE) in EOO
- 253 (Fig. 4 a), followed by snakes ( $\mu$ mean = -22.6  $\pm$  0.25 SE), while almost equal numbers of turtle
- species showed a decline (n = 274) and an increase (n = 205) with decreases being much more
- pronounced than increases ( $\mu$ mean = -17.5  $\pm$  0.72 SE, Fig. 4 e). Almost half of the modeled
- reptile species (n = 3029) showed a strong change in range position, demonstrated by a relatively
- low range overlap (<= 60 %), which was consistent across all three groups (Fig. 4 b,d,f).
- 258 Most of the range centroids (58 %) of all reptile species fell within the Neotropical (n = 1133),
- Afrotropical (n = 1039), Oriental (n = 785) and Australian realm (n = 698). Turtle species had 50
- % of their range centroids in the Nearctic (n = 58), Oriental (n = 53) and Afrotropical realm (n = 58)
- 261 38), while lizards and snakes reflected the overall, total reptile, patterns (Fig. 5 d). Range
- 262 centroids were highly clustered within the different realms, which reflects the overall richness
- hotspots, and hardly any centroids were found in the high northern latitudes (Fig. 5 d). By 2080
- species centroids were projected to shift by a mean distance of 111 km  $\pm$  0.9 SE mostly towards
- a southerly direction. Lizards showed a shift towards all directions, with a slight trend in the
- 266 number of species towards the South (Fig. 5 a), while snakes and turtles show a more
- pronounced shift of species towards the North (Fig. 5 b, c). Turtle ranges shifted by the largest
- distances, followed by snakes (Fig. 5 a-c). The northern realms (Nearctic, Saharo-Arabian,
- 269 Paleartic und Sino-Japanese) showed a dominant shift towards the North, while the southern
- 270 realms (Neotropical, Afrotropical and Australian) showed a dominant shift towards the South.
- 271 The Panamanian, Madagascan and Oriental realms also showed a northerly shift, while the
- Oceanian realm showed a bi-directional shift to the Northwest and Southeast (Fig. 5 d, Fig.
- 273 S2.11). Large realms had a larger proportion of species that shifted their range for a larger
- 274 distance (Fig. S2.11).
- 275 37.4 % of reptile species, for which data would have been available, could not be modeled using
- 276 SDMs, either due to a small sample size or a low model performance (Table 1). We found that
- the species that could not be modeled showed a significant higher mean climatic distance
- 278 between current and future conditions compared to the modeled species and thus occurred in
- areas that experience a stronger change in climatic conditions. This pattern was consistent across
- all three taxa (Fig. 6).
- 281 Looking at the sum of occurrence probabilities, we found similar spatial patterns and a similar
- 282 magnitude in change (Fig. S3.13 S3.15). Projected richness values and their future changes
- were slightly larger under a larger dispersal ability (d/4), but overall all results were consistent
- across the three dispersal scenarios considered (Fig. S4.16 S4.21). Climate change impacts on
- 285 future species richness increased over time, with stronger effects seen for 2080 than 2050, and

the strongest impacts being observed under a high emission scenario (RCP8.5) compared to the

two lower scenarios (Fig. S5.22 - S5.32).

## Discussion

- 289 Reptile richness was projected to decrease significantly across most parts of the world in the
- 290 future (Fig. 1 & 2). This effect was visible across lizards, snakes and turtles alike, although
- regional and species-specific responses differed across the three groups (Fig. S2.7 S2.10).
- 292 Reptile richness is projected to decrease in Brazil, Australia and South Africa, and to increase in
- south-western China and the western United Sates (Fig. 1 e). These areas overlap strongly with
- 294 the biotic convergence zones identified for the conservation of Lepidosaurians (Diele-Viegas *et*
- 295 *al.*, 2020). Brazil in particular is not only characterized by a high reptile richness (Roll *et al.*)
- 296 (2017), Fig. 1 a), but also hosts a large proportion of threatened reptile species (Böhm et al.,
- 297 2013). Furthermore, most protected areas within Brazil are only of low conservation status
- 298 (IUCN Categories V VI) and the stricter ones underrepresent a large proportion of the Brazilian
- 299 biophysical environment (Baldi et al., 2019). While the areas that were projected to show an
- 300 increase in reptile richness with climate warming, i.e. south-west China and the western United
- 301 States, partially overlap with the areas that have the largest proportion of reptile species, they are
- also the ones most threatened by habitat loss from agriculture and logging or harvesting (Böhm
- 303 et al., 2013).
- 304 Species richness changes varied strongly across different regions and different taxa, with lizards
- 305 seeing a strong species increase and decrease in Australia, snakes seeing a strong decrease in
- 306 species richness in South America and turtles seeing a strong increase in the eastern part of the
- 307 United States (Fig. S2.7). All three taxa saw a positive net change in the northern latitudes (Fig.
- 308 S2.7), while lizards saw a decrease in parts of Australia (Fig. S2.7 a,b), snakes in large parts of
- 309 South America (Fig S.2.7 c,d) and turtles saw a strong relative net change in parts of Australia
- and northern Africa (Fig S2.7 f). These differences across groups are also reflected in their
- 311 original richness patterns. Species richness of amphibians, birds and mammals together is a good
- 312 spatial surrogate for species richness of all reptiles combined and of snakes, but not for lizard or
- 313 turtle richness (Roll *et al.*, 2017). Thus, it is not surprising that the areas with the highest decline
- 314 in overall reptile richness (see Fig. 1) strongly overlap with the areas of highest projected
- 315 changes in vertebrate species richness (amphibians, birds and mammals) found by Hof et al.
- 316 (2018), although global reptile richness is largely constrained by temperature, while global
- 317 richness of all other vertebrate groups is mostly constrained by the availability of energy and
- water (Qian, 2010). Historical shifts in geographical ranges and climatic niches further showed
- that niche shifts in endotherms are significantly faster than in ectotherms (Rolland *et al.*, 2018).
- 320 Newbold (2018) further found that Brazil is strongly affected by vertebrate diversity loss due to
- 520 Newbold (2016) further found that Brazil is strongly directed by vertebrate diversity loss due to
- 321 climate change, and together with Australia is also likely to be strongly affected by future land-
- 322 use changes, especially under a high-emission scenario (RCP8.5).
- 323 Globally reptile richness was projected to decline significantly, from about 58 to 53 (9.4 %)
- 324 species on average per grid cell from 1995 to 2080 (Fig. 2 a). This estimate is slightly lower than
- 325 the reptile richness decline predicted by Newbold (2018). One explanation for this difference
- 326 could be that Newbold (2018) only used a subset of species but they also applied a much smaller
- 327 dispersal buffer, which might indicate that our projections provide a rather optimistic scenario.
- 328 Newbold (2018) further found that reptiles together with amphibians are disproportionately

- 329 sensitive to future human land-use. Given the synergistic effect of future climate and land-use
- change on biodiversity (Brook et al., 2008) as well as species population s (Williams et al.,
- 331 2022), land-use change might further exacerbate climate change impacts on global reptile
- 332 distribution and diversity.
- 333 Changes in reptile richness differed among zoogeographic realms, but species richness declined
- 334 significantly across most realms over both time periods (Fig. 2 b,c,d,f,g,h,j,k). Lizards, snakes
- and turtles all showed similar declines in global species richness and across most realms, but
- 336 slightly differed across individual realms. This is in line with a previous study covering various
- realms from tropical to temperate regions which found that 60 % of assessed Lepidosaurian
- species (n = 1114) were vulnerable to changes in climate (Diele-Viegas et al., 2020). We found
- 339 that only the Oceanian and Madagascan realms did not show a consistent decline in species
- 340 richness, which partially corresponds with the results of Diele-Viegas et al. (2020), who found
- that the Madagascan and Oceanian realm were the ones where Lepidosaurians were the least
- 342 vulnerable to climatic change. However, given that the Madagascan realm boasts over 90% of
- 343 endemic reptile species and genera (Glaw & Vences, 2007) and that both realms are completely
- 344 composed of island territories which are usually considered highly vulnerable to climate change
- and might also be affected by future sea level rise and erosion (Diele-Viegas *et al.*, 2020), our
- estimates might in fact underestimate potential climate change impacts in these realms.
- Reptile richness differed strongly across temperature and precipitation, with the highest richness
- being observed in warm, but moist conditions. Under future climate, the climatic conditions with
- 349 high species richness were projected to shift to even more extreme (warmer & wetter) conditions
- 350 (Fig. 3). Reptiles cannot regulate their body temperature internally, so are strongly dependent on
- 351 solar energy captured by the environment to regulate their body temperature (Huey, 1982). This
- 352 might lead to overheating when temperatures reach beyond a species' critical limit, which makes
- 353 them particularly susceptible to climatic changes (Sinervo *et al.*, 2018). However, this might be
- compensated by other biological processes that help species to buffer climate change effects, i.e.
- 355 genomic and phenotypic plasticity (Rodriguez et al., 2017) as well as behavioral and
- 356 physiological adaptation Sunday et al. (2014). Overall, the persistence of reptile species would
- be much more affected by climate cooling than warming, but increasing droughts, which will be
- a consequence of continued warming, have been suggested to pose a significant future threat to
- 359 European reptiles (Araújo *et al.*, 2006). Climate warming will likely have an additional impact
- on reptiles that have temperature-dependent sex determination. Altered sex ratios will not only
- result in a higher extinction risk of local populations, but, together with a reduction in nesting
- 362 sites due to habitat destruction and fragmentation, will also affect the dispersal and potential
- 363 range expansion of a species. Therefore it could also have an impact on population demography
- and size unless temperature shifts in sex determination or female nest-site choice evolves in pace
- with rising temperatures (Boyle *et al.*, 2016; Gibbons *et al.*, 2000).
- 366 The EOO of most species considerably decreased, with lizard species showing the strongest
- decline (Fig. 4 a). Most reptiles further showed a strong decline in range overlap, which was
- 368 consistent across all three groups (Fig. 4). This is in line with results by Warren *et al.* (2018),
- 369 who found that range losses of more than 50% occur in 35% of considered reptile species,
- although this study included only a fraction of all reptile species and dispersal was not
- 371 considered.

- 372 Terrestrial reptiles have narrow niche requirements and small ranges, compared to other
- vertebrate groups, which makes them more susceptible to threats such as habitat loss or invasive
- 374 species (Böhm *et al.*, 2013). 21 % of reptilian species are currently threatened with extinction
- 375 (IUCN, 2022), while adequate baseline data to inform conservation actions is often missing (Roll
- 376 *et al.*, 2017). In addition, cascading effects generated by disease, invasive species, habitat loss
- and climate change might lead to declines of sympatric species and a faster deterioration of
- 378 ecosystem structure than anticipated by climate change alone (Zipkin et al., 2020).
- 379 The majority of reptile species showed a shift towards the South, which was largely driven by
- 380 range shifts in lizards. Turtle ranges overall shifted farthest, followed by snakes (Fig. 5). This is
- 381 likely due to the fact that lizards have the smallest range sizes across the three groups (Roll et al.,
- 382 2017), which in our case also resulted in smaller dispersal distances (Fig. S1.3).
- 383 Range restricted (non-modeled) species are projected to experience significantly higher shift in
- 384 climatic distance than modeled species (Fig. 6), indicating that range-restricted species are
- 385 disproportionately affected by climate change. This highlights once more that sample size
- 386 restrictions of SDMs downplay climate change effects on narrow-ranging and threatened species
- 387 (Platts et al., 2014). Hof et al. (2018) also found significant impacts of climate and land-use
- 388 changes on range-restricted vertebrate species, excluding reptiles. However, similar to the latter
- 389 study we only look at climate anomalies (Euclidean distance between current and future climatic
- 390 conditions) as a metric of climate change, while different metrics have been found to show
- 391 contrasting climate change patterns on a global scale (Garcia et al., 2014). In addition to climate
- 392 change effects, habitat modification has been found to have a more negative effect on small-
- ranging reptile species, as well as species with a small clutch size (Doherty *et al.*, 2020). While
- 394 habitat modification would be an additional factor worthwhile to consider in future impact
- assessments, this would go beyond the scope of this study.
- 396 Our results were strongly dependent on the dispersal, time period and emission scenario (RCP)
- 397 considered, but the overall patterns and richness changes were consistent throughout (see
- 398 Supporting Information). Previous studies have highlighted differences among dispersal, time
- 399 period and RCP across various vertebrate taxa (Thuiller et al., 2019), but also specifically for
- 400 reptiles (Araújo et al., 2006; Newbold, 2018; Warren et al., 2018). Reptile-specific studies have
- 401 either considered no dispersal at all (Araújo et al., 2006; Warren et al., 2018) or a dispersal rate
- 402 of 0.5 km per year (Newbold, 2018). We use a slightly higher biologically informed dispersal
- 403 scenario, with an average of 2-5 km per year, which might be overly optimistic. The considered
- dispersal distances are species-specific and thus strongly depend on the range size of each
- species. Given that our model results as well as the underlying climate scenarios are based on a
- 406 0.5° grid size, small differences in the dispersal distances considered here did not have a strong
- 407 impact on our results (see Appendix S4).
- 408 The projected changes in species distributions help to investigate potential changes in global
- 409 reptile richness patterns and to highlight hotspots of climate change impacts. They also allow to
- 410 compare climate change vulnerability across taxonomic groups. However, these projections are
- 411 far from reality and have to be interpreted with caution. Future studies should try to consider
- 412 additional factors, such as biotic interactions (Davis et al., 1998) and the reshuffling of species
- 413 communities (Voskamp et al., in prep), which might lead to a change in competitive balance
- 414 (Ockendon et al., 2014), altered predator-prey relationships (Harley, 2011) or changes in

- 415 functional diversity (Stewart et al., 2022) and thus the provision of ecosystem functions and
- 416 services (Pecl *et al.*, 2017).

## Conclusion

417

- 418 In addition to climate and land-use changes, reptile species are threatened by habitat loss and
- degradation, invasive species, environmental pollution, disease and biological resource use (e.g.,
- 420 hunting and timber harvesting) (Gibbons *et al.*, 2000, Chapple2021).
- 421 Existing protected areas, sites of biodiversity significance and global conservation schemes do
- 422 not effectively protect reptiles, particularly lizards and turtles (Roll et al., 2017) and 61% of the
- 423 world's skinks do not overlap at all with protected areas (Chapple *et al.*, 2021). The climatic
- 424 space of protected areas is projected to shift considerably with future climate warming (Elsen et
- 425 *al.*, 2020), though even if existing nature reserves would adequately represent reptiles now, they
- 426 might be inadequate to preserve biodiversity in the future.
- 427 Our study shows that reptiles are likely to be considerably impacted by future climate change,
- 428 globally but also within most zoogeographic realms. These impacts are projected to have a
- 429 considerable effect on species' extent of occurrence and range position. Thus, to prevent large
- 430 scale declines in reptile species it is detrimental to lower CO<sub>2</sub> emissions in order to stop on-going
- 431 climate change but also to maintain adequate habitats of sufficient size and quality, especially of
- 432 grassland and savanna habitats (Roll et al., 2017). Furthermore it is necessary to establish new
- 433 protected areas that help to prevent the extinction of particularly vulnerable species, i.e. by
- 434 establishing high-elevation climate refugia within current species ranges (Sinervo et al., 2018).

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# 624 **Data Accessibility Statement**

- 625 GARD range maps are available from https://doi.org/10.5061/dryad.83s7k. EWEMBI and
- 626 ISIMIP2b climate data are available from https://esg.pik-potsdam.de/search/isimip/?
- 627 product=input. The code for the species distribution models can be found at
- 628 https://github.com/xxxx/xxxx, whereas the code and data for the performed data analysis and
- the presented figures, can be found at <a href="https://github.com/xxxx/xxxx">https://github.com/xxxx/xxxx</a>.

# **Tables**

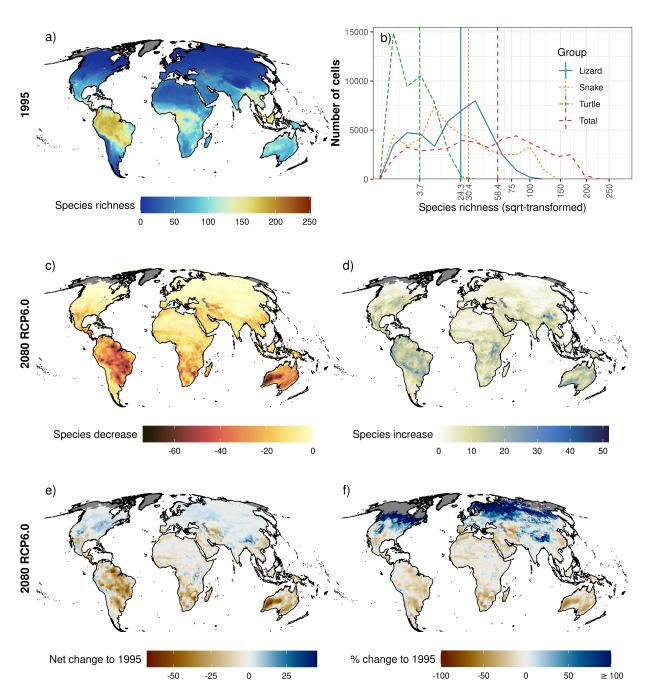
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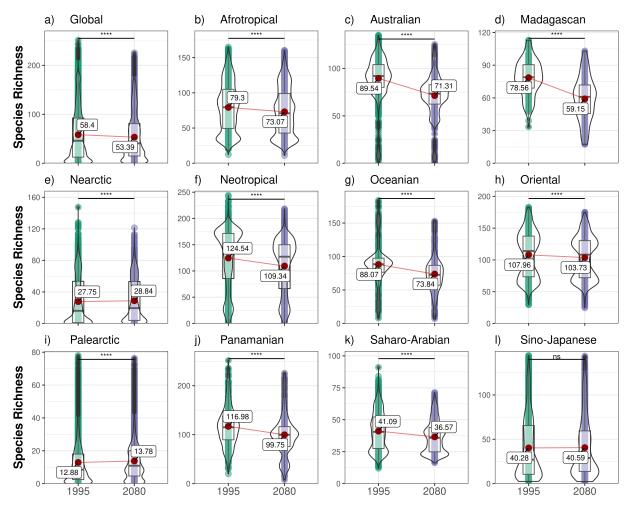
**Table 1.** Number of species that were excluded from the species distribution models due to their restricted range or low model performance.

Taxonomic group	Lizard	Snake	Turtle	Total
No. of species with available data	6328	3414	322	10064
No. of range restricted species (N $\leq$ 10, removed)	2536	1047	19	3602
No. of species with low model performance (AUC < 0.7, removed)	97	62	7	166
Total number of species modeled	3695	2305	296	6296
Percentage of available species modeled	58.4 %	67.5 %	91.9 %	62.6 %

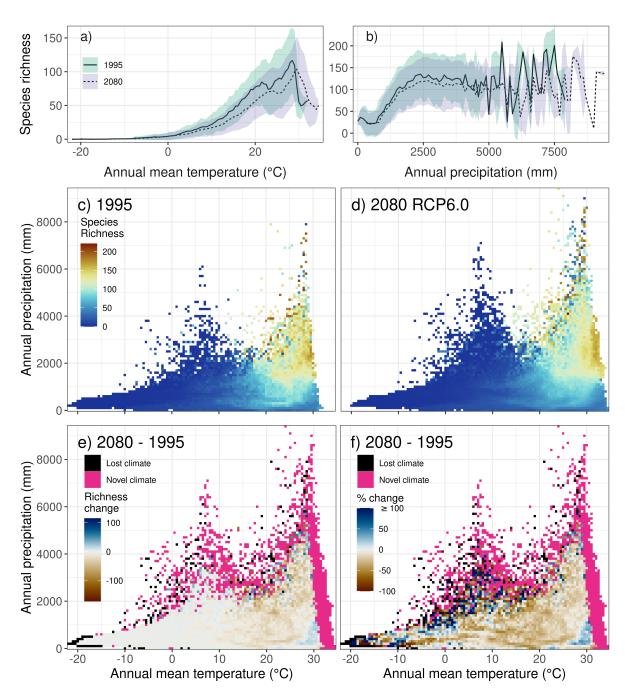
# **Figures**



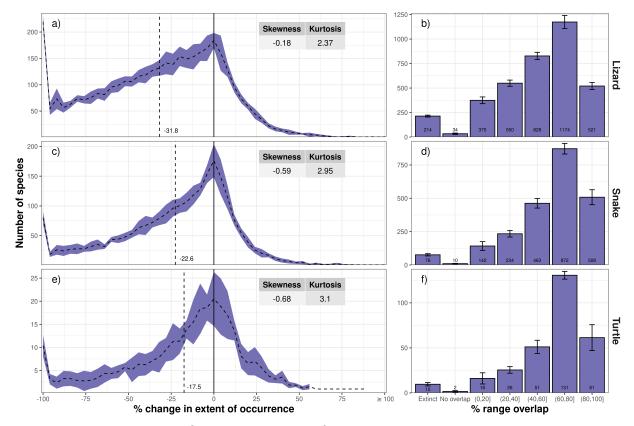
**Figure 1.** a) Map of projected global terrestrial reptile species richness (1995), b) frequency of species richness by taxonomic group (lizard, snake, turtle and total) and c) increase, d) decrease, e) net change and f) relative change (%) in reptile species richness under the representative concentration pathways RCP6.0 and dispersal scenario d/8 for the time period 2080. All maps are in Mollweide equal-area projection (EPSG:54009). Grey areas are regions for which no projections are available. Note that the colour scales differ between the individual panels.



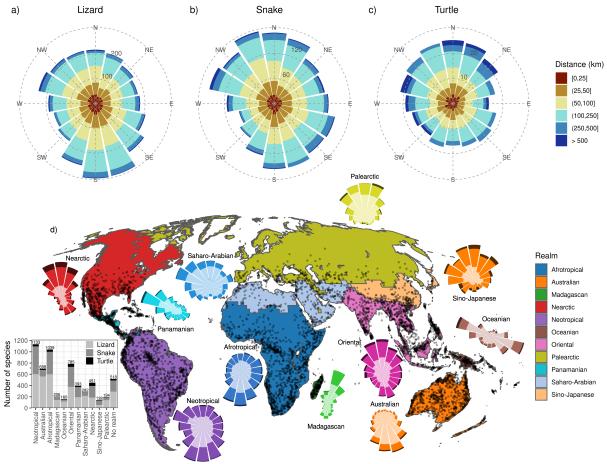
**Figure 2.** Terrestrial reptile species richness across the globe and for each zoogeographic realm (Afrotropical, Australian, Madagascan, Nearctic, Neotropical, Oceanian, Oriental, Palearctic, Panamanian, Saharo-Arabian & Sino-Japanese) over time (1995, 2080) under the representative concentration pathways RCP6.0 and the dispersal scenario d/8. Statistical difference between years was tested using a paired Student's t-test with Holm correction (p < 0.05 = \*, p < 0.01 = \*\*\*, p < 0.001 = \*\*\*\*, p < 0.0001 = \*\*\*\*). Plots show mean (red point & label), median (black horizontal line), 25th to 75th percentiles (box), entire range of data (violin & data points) and density of values (width of violin). Figure 5 provides a map outlining the different zoogeographic realms.



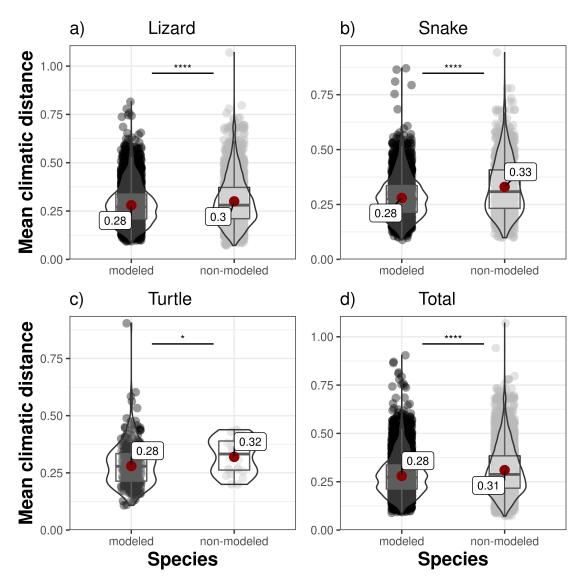
**Figure 3.** Univariate relationship of current (1995) and future (2080 RCP6.0) reptile species richness with a) temperature, b) precipitation and the bivariate relationship of temperature and precipitation with reptile species richness for c) 1995 and d) 2080 RCP6.0 and the respective e) net richness change and f) relative richness change (%) under the dispersal scenario d/8. Lines show the mean and ribbons the standard deviation in variance across space, global circulation model and algorithm.



**Figure 4.** Frequency plots of the mean number of reptile species (a) lizard, b) snake, c) turtle) and their potential future change (%) in extent of occurrence (EOO) and the mean number of reptile species (d) lizard, e) snake, f) turtle) per potential range overlap class (0-20, 20-40, 40-60, 60-80, 80-100). Error margins/bars show standard deviation across the different global circulation models and model algorithms used. Both shown for 2080 under the representative concentration pathway RCP6.0 and the dispersal scenario d/8.



**Figure 5.** Cumulative direction and distance of potential range centroid changes per taxonomic group (b) lizard, c) snake, d) turtle) and d) range centroids (points on map) and the number of species and their directional shift in range centroid position per zoogeographic realm (inset polar plots) for 2080 under the representative concentration pathway RCP6.0 and the dispersal scenario d/8. Please note that some range centroids did not fall inside the zoogeographic realm boundaries and thus were not associated with an realm (n = 477).



**Figure 6.** Mean climatic distance for modeled and non-modeled species, split by taxonomic group (a) lizard, b) snake, c) turtle and d) total). Statistical difference between modeled and non-modeled species was tested using a Student's t-test with Holm correction (p < 0.05 = \*, p < 0.01 = \*\*\*, p < 0.001 = \*\*\*\*, p < 0.0001 = \*\*\*\*). Plots show mean (red point & label), median (black horizontal line), 25th to 75th percentiles (box), entire range of data (violin & data points) and density of values (width of violin). Results are shown for 2080 under the representative concentration pathway RCP6.0.