# Global determinants of zoogeographical boundaries

2 Gentile Francesco Ficetola<sup>1,2</sup>\*, Florent Mazel<sup>1,3</sup> and Wilfried Thuiller<sup>1</sup> 3 <sup>1</sup> Univ. Grenoble Alpes, CNRS, Laboratoire d'Écologie Alpine (LECA), F-38000 Grenoble, France 4 <sup>2</sup> Department of Biosciences, Università degli Studi di Milano. Via Celoria 26, 20133 Milano, 5 6 Italy <sup>3</sup> Department of Biological Sciences, Simon Fraser University, Burnaby, BC, Canada 7 \*e-mail: francesco.ficetola@gmail.com 8 9 10 Acknowledgements We thank S. Ramdhani for providing high-resolution maps of bioregions. The research leading to 11 these results has received funding from the European Research Council under the European 12 Community's Seven Framework Programme FP7/2007-2013 Grant Agreement no. 281422 13 (TEEMBIO). All authors belong to the Laboratoire d'Écologie Alpine, which is part of Labex 14 15 OSUG@2020 (ANR10 LABX56). 16 17 Author contributions 18 The three authors conceived the study. GFF performed all analyses with the help of FM. GFF wrote the first version of the manuscript, with contribution of all authors 19 20 Competing interests 21 The authors declare no competing financial interests 22 23 24 Data Deposition statement The data and the scripts that support the findings of this study are available from the 25 corresponding author upon request. 26

The distribution of living organisms on Earth is spatially structured. Early 27 biogeographers already identified the existence of multiple zoogeographical regions, 28 characterized by faunas with homogeneous composition that are separated by 29 biogeographical boundaries. Yet, no study has deciphered the factors shaping the 30 distributions of terrestrial biogeographical boundaries at the global scale. Here, using 31 spatial regression analyses, we show that tectonic movements, sharp changes in climatic 32 conditions, and orographic barriers determine extant biogeographical boundaries. 33 These factors lead to abrupt zoogeographical transitions when they act in concert, but 34 their prominence varies across the globe. Clear differences exist among boundaries 35 representing profound or shallow dissimilarities between faunas. Boundaries separating 36 zoogeographical regions with limited divergence occur in areas with abrupt climatic 37 transitions. On the other hand, plate tectonics determine the separation between deeply 38 39 divergent biogeographical realms, particularly in the Old World. Our study reveals the multiple drivers that have shaped the biogeographical regions of the world. 40

### Background

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Naturalists have long been fascinated by the variation of life across geographical regions, and have described biogeographic areas since the 18th century <sup>1-5</sup>. Wallace <sup>4</sup> was one of the first mapping these biogeographical regions, and identified some areas of transition between them (biogeographical boundaries). The analysis of biogeographical patterns has since remained an active research field <sup>6-8</sup> and, in the last years, the increasing availability of species distribution data has fostered quantitative studies on biogeographical regionalization at both global and regional scales, using macroecological and geospatial approaches 9-15. On one hand, several biogeographical regions are clearly separated by barriers to dispersal <sup>16</sup>. For instance, Australia and Madagascar have unique terrestrial faunas, and their distinctiveness clearly derive from the fact that they remained isolated from other land masses for tens of million years. On the other hand, many delineated biogeographical boundaries cross continents or correspond to narrow sea straits (Fig. 1). These terrestrial boundaries are assumed to be the consequence of multiple factors limiting the interchanges across regions, such as the presence of unfavourable climates, high turnover of environmental conditions, orographic barriers and historical geological and climatic isolation <sup>7,16,17</sup>. Despite those qualitative statements, we do not know much about the relative importance of those determinants to delineate biogeographical boundaries <sup>18</sup>, and no formal and comprehensive analyses have been carried so far. Until now, studies on biogeographical boundaries generally focused on one specific area, such as the Wallace line or the Nearctic-Neotropical transition zone <sup>16,17</sup>, while a global analysis is still lacking.

We believe that this lack of knowledge comes from the complex nature and definition of biogeographical boundaries. Indeed, there is no single definition of boundary and they appear to be hierarchically structured and spatially heterogeneous. For instance, Holt et al. 11 recently delineated the zoogeographical regions of the world by integrating species distribution data of terrestrial vertebrates with phylogenetic information. Measuring the phylogenetic turnover between vertebrate assemblages (taken at 200 × 200km resolution) and using a cluster algorithm, they delineated twenty zoogeographical regions of the world that explain most of variation in biodiversity while maximising the phylogenetic dissimilarities between them <sup>11</sup>. Interestingly, the nested nature of the dendrogram created from their cluster analysis also allowed Holt et al. to identify eleven regions, at a higher level, called realms (Fig. 1) 11. However, the position of cut-off points is somehow arbitrary and, along the same dendrogram, if a deeper cut-off of similarity is used, some of Holt's realms collapse, resulting in a smaller number of realms that are mostly consistent with the original maps of Wallace's realms <sup>19</sup> (Fig. 1b). In other words, some boundaries separate highly dissimilar assemblages, while others separate regions with lower dissimilarities (Fig. 1). To refer to this biogeographical hierarchy, and since there is no clear terminology yet, we will use the terms shallow, intermediate and deep bioregions and boundaries. Clearly, complex determinants are responsible for this nested structure of biogeographical regions and we argue that some might explain deep bioregion boundaries, while others should be more related to intermediate and shallow boundaries. More specifically, we hypothesise that 1) climatic heterogeneity, 2) orographic barriers, 3) past tectonic history and 4) velocity of past climate change may play a major role in setting biogeographical boundaries. These factors may have a different role in explaining shallow or deep boundaries, as processes acting deeper in the past (e.g. plate tectonic movements) may be most important for deep boundaries, while factors representing present-day ecological barriers (e.g. climatic heterogeneity) may best explain shallow boundaries.

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Climate is a major determinant of the present-day limits of species distributions <sup>20</sup>, and faunistic turnover is higher between regions with dissimilar environmental features <sup>21,22</sup>. Therefore, climate could have a major role, for instance for shallow boundaries <sup>18</sup>. However, climatic conditions have strongly shifted during the Quaternary, determining broad scale changes of species distributions and modifications of assemblages <sup>23-25</sup>. The velocity of past climate change since the last glacial maximum is known to be a major driver of endemism and biogeographical structure, with higher endemism of vertebrates in regions with more stable climate <sup>26</sup>. As endemism plays an important role in the definition of biogeographical regions <sup>19</sup>, quaternary climate changes have been potentially important to set boundaries representing shallow or intermediate dissimilarity among regions <sup>23</sup>. Tectonics have determined the long-term isolation of the biotas on some continental plates <sup>16</sup>, thus we expect that tectonic history (movements of plates during the Cenozoic) has determined some of the deepest boundaries <sup>7,27</sup>. While the role of tectonics on biogeographical patterns has long been recognized <sup>16</sup>, no global study has used plate-motion models to explicitly quantify determinants of biogeographical boundaries. Finally, mountains are major barriers to dispersal of terrestrial animals, thus we expect an overall role of orographic barriers.

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Here, we build on Holt's zoogeographical regionalization by quantitatively measuring the relative importance of the above-mentioned four hypotheses across the nested structure of the global regions. First, we used spatial regression models to identify the factors best explaining the occurrence of boundaries. Second, we mapped their spatial heterogeneity, to identify global and regional variation of processes in function of climate and geological history. Third, we explored their relative importance through the nested structure of regions, to assess whether these processes play a consistent role on all the boundaries, or whether some are more important for boundaries representing deep or shallow dissimilarity. Finally,

we demonstrated the robustness of our conclusions to alternative classifications of zoogeographical regions <sup>6,10</sup>.

#### Results

The geographical position of terrestrial biogeographical boundaries was accurately predicted by the spatial models (Supplementary Table 1). When we analysed the factors related to the overall presence of boundaries (all boundaries in Fig. 1), we found support for a joint role of climatic heterogeneity, tectonic movements during the last 65 million years, and orographic barriers (Fig. 2, Supplementary Table 1). Temperature heterogeneity and tectonic movements were the variables with the strongest overall effect size, followed by orographic barriers and heterogeneity of temperature seasonality. We did not detect any relationship between biogeographical boundaries and the velocity of late quaternary climate change. Velocity of climate change is strongly related to topography  $^{26}$  (Supplementary Table 2), still it remained non-significant if altitude was excluded from the model (simultaneous autoregressive model:  $t_{2191} = -0.73$ , P = 0.46).

Geographically weighted regression (GWR) suggested that relationships between environmental features and boundaries were not homogeneous across the globe (Fig. 3a-d). Overall, temperature heterogeneity best explained the boundaries crossing Eastern Asia, Central and North America, while heterogeneity of temperature seasonality best explained the boundaries of the Amazonian and Guineo-Congolian regions. Western Eurasia boundaries were best explained by tectonic movements while orographic barriers best explained the Asiatic boundaries between the Arctico-Sibirian, the Eurasian, the Tibetan and the Oriental regions (Fig. 4a). Climatic variables were particularly important to define the boundaries of

tropical and subtropical regions. Species turnover is the basis of biogeographical regionalization, and is more strongly linked to environmental heterogeneity in the tropics than at the high latitudes <sup>21</sup>. This probably occurs because the limited short-term climatic variability in the tropics can favour physiological specialization, determining narrower niches and particularly strong responses to climate <sup>28</sup>.

We then performed sequential analyses on boundaries representing different levels of faunistic dissimilarities. The boundaries representing the shallowest dissimilarities (white lines in Fig. 1) were strongly associated to heterogeneity of temperature seasonality and, to a lesser extent, to orographic barriers (Fig. 2, Supplementary Fig. 1). Major equatorial regions (Guineao-Congolian and Amazonian) are areas with constant temperature through the year (Supplementary Fig. 2) and their limits, particularly in the south, are strongly related to shifts toward more seasonal climates. This strongly agrees with the idea that limited seasonal variability is a major determinant of the narrow niche of tropical animals <sup>28</sup>.

When we focused on deeper biogeographical relationships (intermediate bioregions, i.e. boundaries among Holt's realms), heterogeneity of temperature was the variable with the strongest effect size, followed by plate tectonic movements and orographic barriers (Fig. 2, Supplementary Fig. 1, Supplementary Table 1). Finally, the deepest biogeographical boundaries were mostly related to plate tectonic motion, with a consistent effect through the boundaries crossing the whole Old World (Figs. 2-4, Supplementary Table 1). Nevertheless, significant local relationships remained with climatic parameters and orographic barriers (Fig. 3), and the position of the boundary between the Neotropics and the Nearctic corresponded to areas with strong heterogeneity of temperature (Fig. 3e, Fig. 4b). The optimal bandwidth detected by geographically weighted regressions was 1000 km in the analysis of shallow boundaries, 1800 km when focusing on the intermediate boundaries, and 4800 km for deep boundaries. In these spatial regression models, the optimal bandwidth identifies the distance

of neighbours to include into local regressions <sup>29</sup>, and the shorter bandwidths of shallow and intermediate bioregions suggest that more local processes act on the boundaries representing limited dissimilarities.

#### **Discussion**

Our analysis is a first attempt to tease apart the role of multiple factors in shaping zoogeographical boundaries at the global scale, and it shows that multiple factors often interplay to determine major transitions. For instance, past separation of tectonic plates led to long term isolation and strong dissimilarity of faunas among continents, but biotic interchanges have occurred when the movement of some plates brought isolated biotas in contact <sup>30-32</sup>. Clear biogeographical differences have remained even after the contact among plates, likely maintained by the interplay with other processes. In the Old World, the collision between the African, the Arabian, the Eurasian and the Indian plates has created major mountain chains, which are physical barriers that also determine sharp climatic transitions (Supplementary Fig. 4). In this region, plate tectonics, climate and orography have thus played a joint, and difficult to disentangle, role in shaping zoogeographical boundaries (Fig. 3).

Conversely, no sharp barriers exist between the Neotropics and the Nearctic, thus the transition between these two realms is more blurred <sup>7,19,33</sup>. The northern distribution limit of Neotropical taxa is highly heterogeneous, with some Neotropical families of vertebrates limited to areas south of Panama, and others ranging until Texas <sup>16</sup>. The formation of the Panama Isthmus was a complex geological process, with multiple waves of dispersal of terrestrial organisms <sup>32,34</sup>, and the deepest present-day faunistic transition does not always

coincide with the narrowest isthmus or with the point of contact between plates (Uramita Suture) <sup>16,22,34</sup>. The dispersal of organisms between North and South America was likely limited by the interplay between availability of land and suitable environmental conditions <sup>32,34</sup>, and the transition from tropical to more temperate climates remains the most likely factor limiting biotic homogenisation (Figs. 3 & 4). A long standing debate exists on the boundaries of some regions, such as the position of the southern limit of the Nearctic, or the existence of the boundaries of the Sino-Japanese region, and some of them have been proposed as possible transition zones <sup>19,35</sup>, even though they harbour many endemic taxa and maintain distinct biotas <sup>16,36</sup>. Temperature heterogeneity is the strongest correlate of the boundaries of these regions (Figs 3 & 4). Climatic, tectonic and orographic changes are often closely linked, but our results suggest that complex faunistic transitions may be associated to areas where climate does not act jointly with other processes.

The boundaries across Eurasia (e.g. between the Palearctic and the Saharan region, between the Sino-Japanese and the Oriental regions) were strongly related to tectonic movements, i.e. the recent contact between the Eurasian, the Arabian and the Indian plate <sup>37</sup>, a pattern well recognized in the biogeographical literature <sup>16,38,39</sup>. The importance of tectonic movements was particularly clear in Western Asia (Fig. 3c). In this region, the boundary between the Saharan and the Eurasian bioregions matches well the limits of the Arabian plate, which remained isolated from Eurasia until the Miocene <sup>37,38</sup>. The formation of major mountain chains (e.g. the Zagros Mountains) after the collision between Arabia and Eurasia, and the harsh climatic conditions, have probably contributed to the strong differentiation between the Arabian and the Eurasian faunas <sup>16</sup>. The GWR analysis performed on all boundaries taken together suggested that tectonic movements have a very broad influence over Western Eurasia, with apparent effects spanning northward up to the Urals (Fig. 3c). However, this is likely an artefact of GWR analysis, which, in this case, overestimated the

influence of tectonics across space, probably because of the very strong local effect of the movements of the Arabian plate. There is indeed no global effect of tectonics on shallow boundaries (such as the one between the Eurasian and the Arctico-Siberian plates; Fig. 2). Furthermore, no tectonic movements occurred inside the Eurasian plate during the last 100 million years <sup>37</sup> (Supplementary Fig. 4), and the boundary between the Eurasian and the Arctico-Siberian plate was clearly unrelated to tectonic movements if analysed separately (Supplementary Fig. 1).

Boundaries in Eastern Asia and between the bioregions of central-northern America were related to the presence of a strong temperature gradient (Fig. 3a). Regional scale analyses on Eastern Asia yielded a similar pattern, and showed that the interplay between present-day climate and elevational gradients is a strong determinant of zoogeographical boundaries in this area <sup>39</sup>. He, et al. <sup>39</sup> suggested that orographic barriers and tectonics were the most likely determinant of biogeographical structure in Western China, while the transition from tropical to temperate and continental climates was a major determinant of the regionalization in Eastern China <sup>39</sup>, which corroborate our findings.

Here we focused on the biogeographical boundaries proposed by Holt et al <sup>11</sup>. Alternative biogeographical structures have been proposed using both qualitative and quantitative approaches <sup>6,10,12-14,16</sup>. Although some differences exist, the overall pattern is consistent among studies, and differences are mostly for the shallow boundaries between subregions, while the deepest boundaries are strikingly similar between Wallace's <sup>4</sup> original classification and modern, data-demanding approaches. Interestingly, these boundaries that remain highly congruent among studies are the ones we showed that arise from several factors, such as the joint effect of tectonics, climate and orography in the Old World (Fig. 3fg). Actually, our conclusions on how multiple processes act in concert to define the deepest biogeographical dissimilarities are robust, and do not strongly change if we use alternative

regionalizations <sup>6,10</sup> as baselines (Supplementary Table 3, Supplementary Fig. 3, Supplementary Discussion). The situation is more complex for boundaries representing shallow dissimilarities, which may be blurred by the presence of transition zones <sup>13</sup> and for which different taxa can show non-congruent regionalization <sup>10-12</sup>. Furthermore, responses to climatic factors may be strongly different among taxa, meaning that the parameters determining boundaries may vary not only among areas of the world, but also depending on the taxa on which biogeographical analyses are based. Fine resolution analyses, focusing on specific boundaries, can be important to reveal additional processes acting at more regional scale, and to understand when the biogeographical structure has originated <sup>18,33,40,41</sup>. Nevertheless, the analysis presented here paves the way for in-depth examination and comparative tests of the factors driving ecological and biogeographical transitions at multiple scales and for multiple taxa. The zoogeographical regions of the world have been shaped by multiple ecological and historical drivers. Using adequate spatial models, in combination with well-defined factors representing ecological expectations, allows to identify the complex and hierarchical processes determining zoogeographical boundaries, thus enabling a more objective understanding of biogeographical patterns.

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## Methods

Data

Biogeographical regions – We built on Holt's maps of biogeographical regions <sup>11</sup> that we
converted in a raster grid at a 200 km resolution (Mollweide equal-area projection; see
Supplementary Figs. 2 and 4 for Earth maps at this resolution), a scale generally appropriate
for global analyses of species distribution <sup>42,43</sup> . The "terrestrial" biogeographical boundaries
were defined as the boundaries between zoogeographical regions that are not separated by the
sea at this resolution (Fig. 1). A cell was considered to be on the boundary if a nearby cell
belongs to a different zoogeographical region / realm (depending on the analysis). A few
boundaries were represented by narrow sea straits, that are not evident at the 200 km
resolution (Gibraltar, Djibouti and La Pérouse Straits; see Fig. 1 and Supplementary Fig 2),
and were also considered among the analysed boundaries.
Predictors - We considered four processes that might be related to the probability that a given
world cell represents biogeographical boundaries: 1) areas of high climatic heterogeneity
(climatic barriers); 2) orographic barriers; 3) tectonic separation; 4) instability of past climate.
The climatic heterogeneity hypothesis proposes that boundaries correspond to areas where
climatic parameters show strong spatial turnover (heterogeneity among neighbouring cells).
We considered the heterogeneity for four climatic variables: annual mean absolute
temperature, temperature seasonality, annual summed precipitation and precipitation
seasonality; all climatic variables were extracted from the Worldclim dataset 44 up-scaled at a
200 km resolution. These variables represent both average conditions and their variability
across the year, and are simple major determinants of vertebrate distribution <sup>45</sup> . Furthermore,
mean annual temperature and precipitation seasonality are enough to explain most of climatic
variation at the global scale <sup>21</sup> , and other important variables (e.g. summer and winter

temperatures) are strongly related to linear combinations of the four climatic parameters considered in our analyses (Supplementary Table 4). To measure local heterogeneity, for each cell, we calculated the coefficient of variation (CV) between the focal cell and the neighbouring ones, using a queen connection scheme. Therefore, the values at a given cell are higher if the cell is strongly different from the neighbours (Supplementary Fig. 4). To test for the orographic barrier hypothesis, we calculated the mean absolute difference between the altitude of each cell and the neighbouring ones. To test for the potential effect of past climatic change/stability, for each cell we calculated the average velocity of climate change since the last glacial maximum <sup>26</sup>. Past climate change from the Cenozoic could also likely explain present-day biogeographical structure. However, given that paleoclimatic reconstructions are still unable to reliably reproduce deep past climates 46-48, we preferred not including them in our analyses. To test for the tectonic separation hypothesis <sup>7</sup>, we calculated the variability in geographical distance between each cell and its neighbours during the last 65 million years (i.e. temporal variability of geographical distances averaged across neighbours; see Supplementary Fig. 4 for details and examples) using the GPLATE software <sup>49,50</sup>. This value is low for cells that did not change their position compared to neighbours (e.g. within continental shelfs) and increases for cells that experienced tectonic movements (e.g. a continental collision) (Supplementary Figs. 4-5). All variables were log-transformed prior to analyses to improve normality and reduce skewness; pairwise correlations between the seven variables were < 0.7; the strongest correlations were between mean temperature heterogeneity and altitude variation, and between velocity of past climate change and altitude variation (Supplementary Table 2).

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#### Statistical analyses

We used spatially-explicit regression models to assess the factors that may explain the position of biogeographical boundaries. We first analysed the factors related to the overall presence of boundaries (all boundaries in Fig. 1, global analysis). The dependent variable was whether a grid cell is in contact with a terrestrial biogeographical boundary (Y/N; Fig. 1), while the seven environmental variables, scaled to mean = 0 and variance = 1, were the independent ones. We then performed three analyses, to assess the factors related to boundaries representing different values of phylogenetic turnover: shallow phylogenetic turnover (boundaries between shallow bioregions but not between realms; white lines in Fig. 1), deep turnover (boundaries between intermediate and deep bioregions, i.e. Holt's realms) and very deep turnover (boundaries between deep bioregions, i.e. Wallace's realms). These analyses were performed to assess the relative importance of variables identified by the global analysis in determining boundaries representing specific levels of turnover, thus we used variables significant in the global analysis as independent. Each analysis was limited to within 1000 km from the target biogeographical boundaries, to avoid an excessive number of zeros.

The residuals of preliminary ordinary least squares regression showed significant spatial autocorrelation (global analysis: Moran's I=0.357; analysis on shallow boundaries: I=0.374; analysis on intermediate boundaries: I=0.361; analysis on deep boundaries: I=0.366; all P<0.001), and failure in taking into account spatial autocorrelation may bias the result of regression analyses  $^{51}$ . Therefore, we used simultaneous autoregressive spatial models (SAR) with binomial error distribution to identify the environmental features related to the occurrence of biogeographical boundaries. SAR-models are spatially-explicit regression techniques that deal with spatial autocorrelation; in our models, spatial autocorrelation was incorporated in the error term using neighbourhood matrices (SAR<sub>ERR</sub>). SAR<sub>ERR</sub> is considered among the best-performing approaches to spatial regression  $^{51-53}$ . We used a neighbourhood of 566 km, which is the shortest distance allowing to keep all study cells connected to at least

another cell. Binomial SAR<sub>ERR</sub> were built using hierarchical generalized linear mixed models (HGLM) with spatially correlated random effects <sup>54</sup>. HGLM provide results consistent with other analytical approaches, e.g. spatial mixed models <sup>55</sup>, but are more computationally efficient, allowing to analyse large datasets in reasonable time <sup>54</sup>. In all models, variance inflation factor was ≤ 3 for all variables, indicating that collinearity among variables was not a major issue <sup>56</sup>. Nevertheless, moderate correlation existed between altitude variation and mean temperature heterogeneity (Supplementary Table 2). We thus repeated analyses by removing the correlated variables; coefficients obtained removing the correlated variables were in good agreement with the ones of the full models (Supplementary Table 5), confirming the robustness of our analyses. Analyses were performed on the R environment with the packages car, hglm, maptools, raster and spdep <sup>57-60</sup>. The capability of SAR models to correctly predict the position of biogeographical boundaries was assessed using the maximum true skill statistics, which is a measure of predictive accuracy ranging from -1 to +1, where +1 indicates perfect agreement between observed and predicted values, and values ≤ 0 indicate that performance is not better than random <sup>61</sup>.

SAR models provide one single coefficient per each independent variable, representing the overall relationship (global analysis), but biogeographical and ecological relationships can often vary as a function of the location, showing strong spatial heterogeneity <sup>62</sup>. We thus used geographically weighted regression (GWR) to assess the spatial heterogeneity of relationships between environmental features and boundaries. GWR is an exploratory technique that pinpoints where non-stationarity occurs within the geographical space, i.e. where locally-weighted regression coefficients deviate from their global values. If the local coefficients vary across space, this may be considered as an indication of non-stationarity <sup>29</sup>. GWR was performed after the SAR<sub>ERR</sub> analyses, considering variables significant in SAR<sub>ERR</sub>. We used a binomial model and standardized independent variables.

The best bandwidth was identified through a fixed Gaussian Kernel; to identify the best bandwidth, we built all the models with bandwidths from 5000 to 1000 km at intervals of 200 km, and selected the one with lowest corrected Akaike's Information Criterion (AICc). GWR was run using the software GWR4.0.80 <sup>63</sup>; Local significance of GWR was adjusted for multiple testing following <sup>64</sup>.

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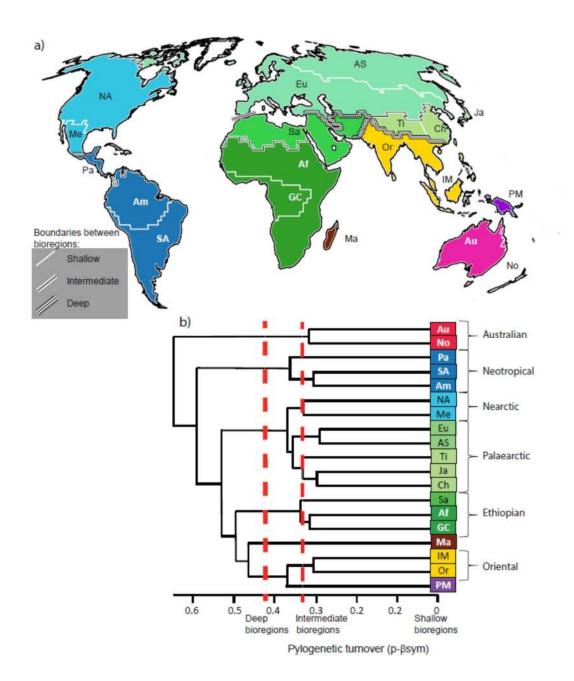
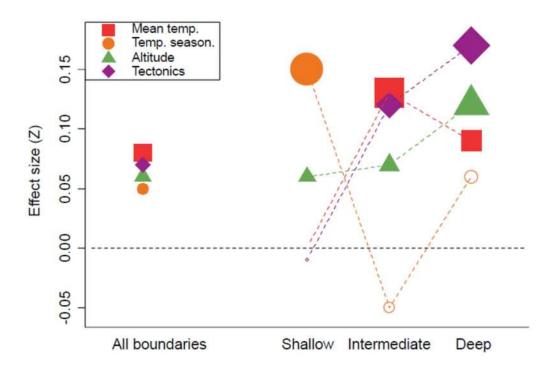


Figure 1. The global zoogeographical regions of the world, as defined by Holt et al. <sup>11</sup>. a) Biogeographical regions for vertebrates and their associated boundaries used here, as defined on the basis of phylogenetic faunistic turnover <sup>11</sup>. b) Phylogenetic turnover (p-βsim; <sup>11</sup>) among bioregions. Regions may be clustered at different turnover thresholds. Clustering them at p-βsim = 0.33 results in bioregions corresponding to the Holt's realms <sup>11</sup>, while clustering them at deeper p-βsim values results in bioregions very similar to the traditional biogeographical realms <sup>6,19</sup>. The figure has been redrawn on the basis of Holt et al. <sup>11</sup>. Biogeographical regions are: Au, Australian; No, Novozelandic; Pa, Panamian; SA, South American; Am, Amazonian; NA, North American (=Nearctic); Me, Mexican; Eu, Eurasian; AS, Arctico-Siberian; Ti, Tibetan; Ja, Japanese; Ch, Chinese; Sa, Saharo-Arabian; Af, African; GC, Guineo-Congolian; Ma, Madagascan; IM, Indo-Malayan; Or, Oriental; PM, Papua-Melanesian. The Polynesian region is not shown.



**Figure 2.** Relative importance of plate tectonics, altitude and climate on the biogeographical regions boundaries position worldwide. The figure presents the effect sizes (obtained through autoregressive models) of each factor in explaining all boundaries, and boundaries between shallow, intermediate and deep bioregions (19, 11 and 6 bioregions, respectively). The size of symbols is proportional to effect size; empty symbols represent non-significant values. Effect size was measured using Fisher's Z, which allows the comparison among analyses even though they have different sample size <sup>65</sup>.

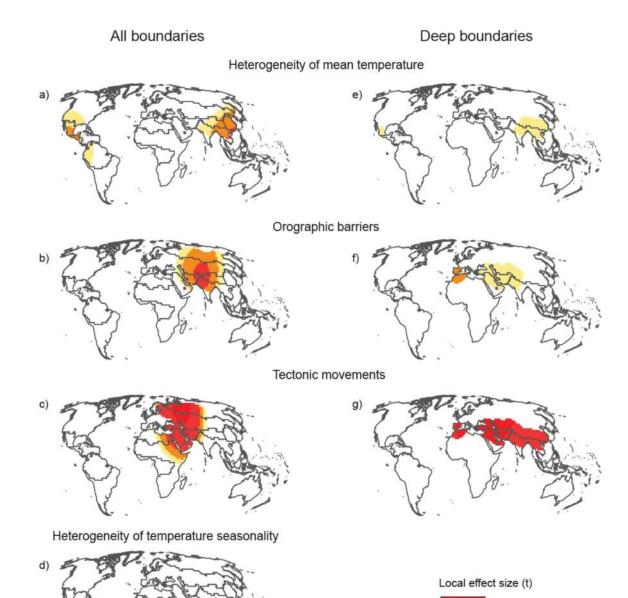


Figure 3. The geographical variability of the importance of tectonics, altitude and climate on the position of biogeographical boundaries: heterogeneity of local effect sizes obtained through geographically weighted regression. Left-panels: analysis on all the boundaries; right-panels: analysis limited to the deep boundaries. Only local effect sizes significantly higher than zero are mapped. See Supplementary Fig. 1 for the results of analyses on shallow and intermediate boundaries.

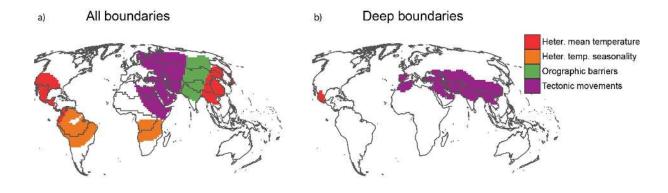


Figure 4. Factors most strongly related to the presence of biogeographical boundaries. For each pixel, the map shows the factor with the highest local effect size according to geographically weighted regression. Only local effect sizes significantly higher than zero are mapped.